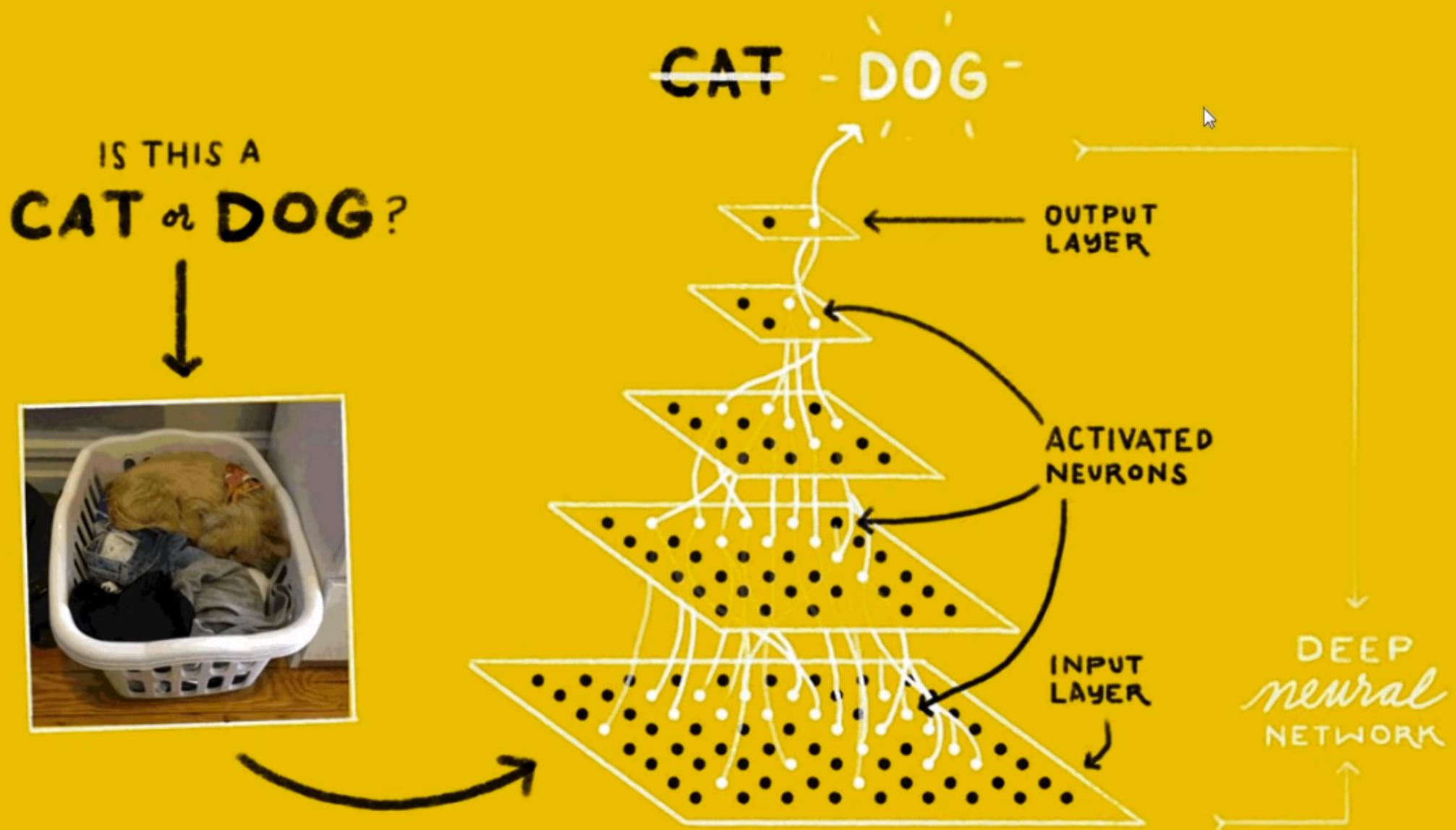
Visualizing Dataflow Graphs of Deep Learning Models in TensorFlow

Kanit "Ham" Wongsuphasawat @kanitw University of Washington

Daniel Smilkov, James Wexler, Jimbo Wilson, Dandelion Mané, Doug Fritz, Dilip Krishnan, Fernanda B. Viégas, Martin Wattenberg Google Research







From "Large-Scale Deep Learning with TensorFlow," Jeff Dean - https://youtu.be/vzoe2G5g-w4

Develop

API r1.3

Deploy

Extend

An open-source software library for Machine Intelligence

GET STARTED





TensorFlow 1.3 has arrived!

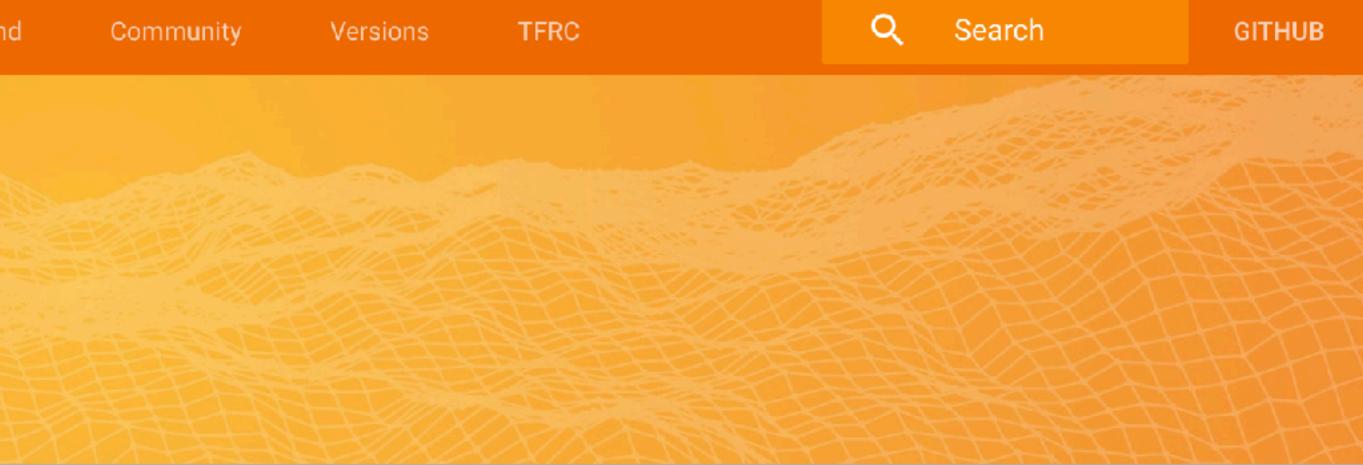
We're excited to announce the release of TensorFlow 1.3! Check out the release notes for all the latest.

Introducing TensorFlow Research Cloud

We're making 1,000 Cloud TPUs available for free to accelerate open machine learning research.

LEARN MORE

UPGRADE NOW





The 2017 TensorFlow Dev Summit

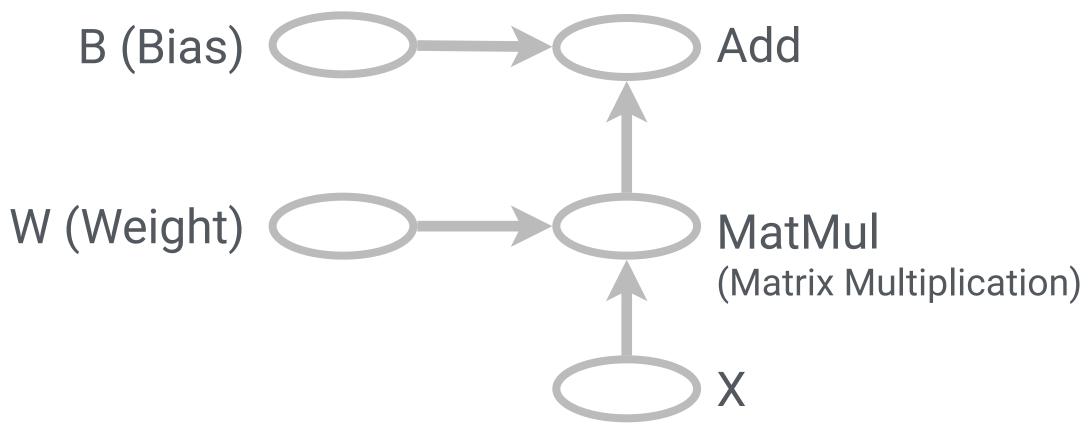
Thousands of people from the TensorFlow community participated in the first flagship event. Watch the keynote and talks.

WATCH VIDEOS



TensorFlow: Dataflow Graphs for Machine Learning

Using dataflow graphs to represent computation.



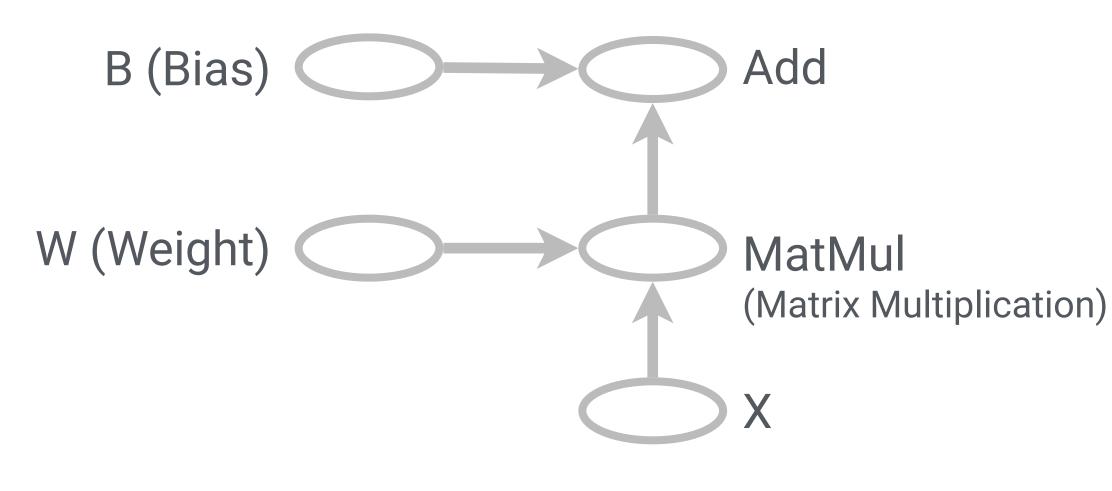
Dataflow graph for a linear model (WX+B)





TensorFlow: Dataflow Graphs for Machine Learning

Using dataflow graphs to represent computation.



Dataflow graph for a linear model (WX+B)

nodes = operations

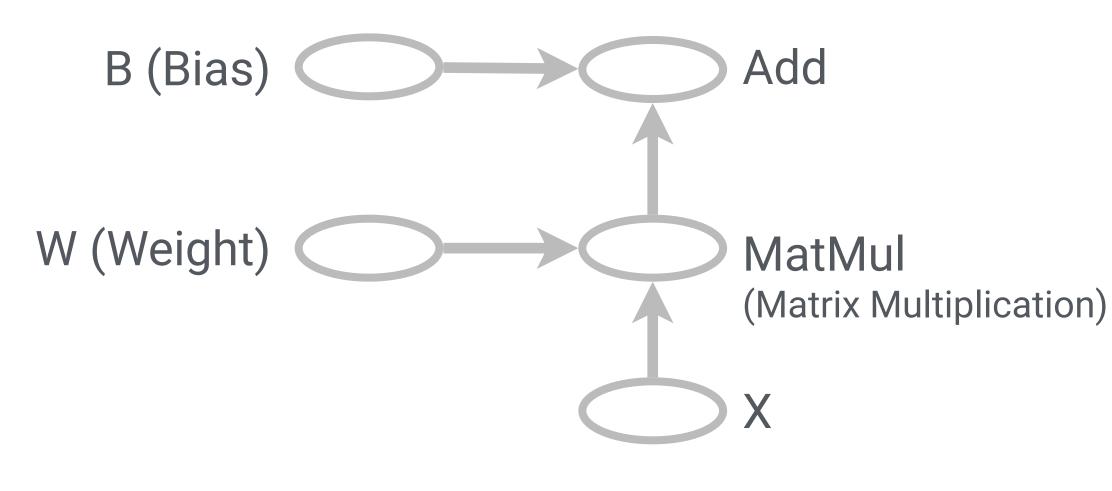
e.g., mathematical functions, constants (initializing values), summaries (logging data)





TensorFlow: Dataflow Graphs for Machine Learning

Using dataflow graphs to represent computation.



Dataflow graph for a linear model (WX+B)

modes = operations

e.g., mathematical functions, constants (initializing values), summaries (logging data)

edges = data

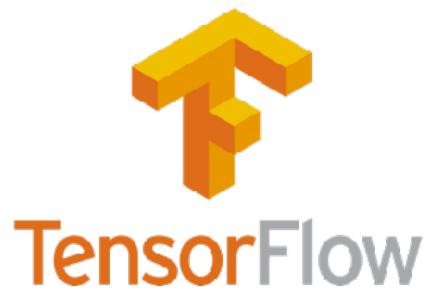
multi-dimensional array (tensors)

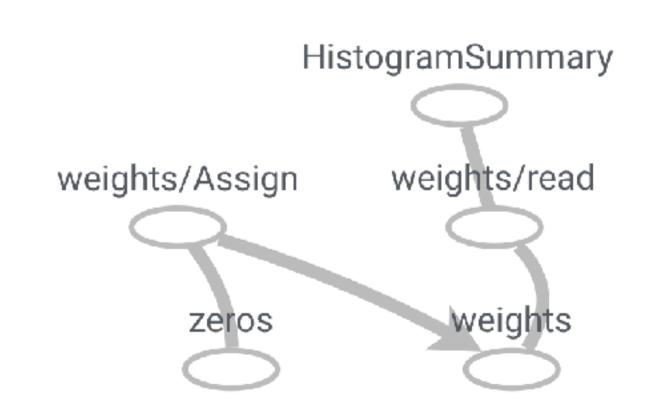




```
zeros = tf.zeros([784, 10])
W = tf.Variable(zeros, name='weights')
tf.histogram_summary('weights', W)
```

High-level Program

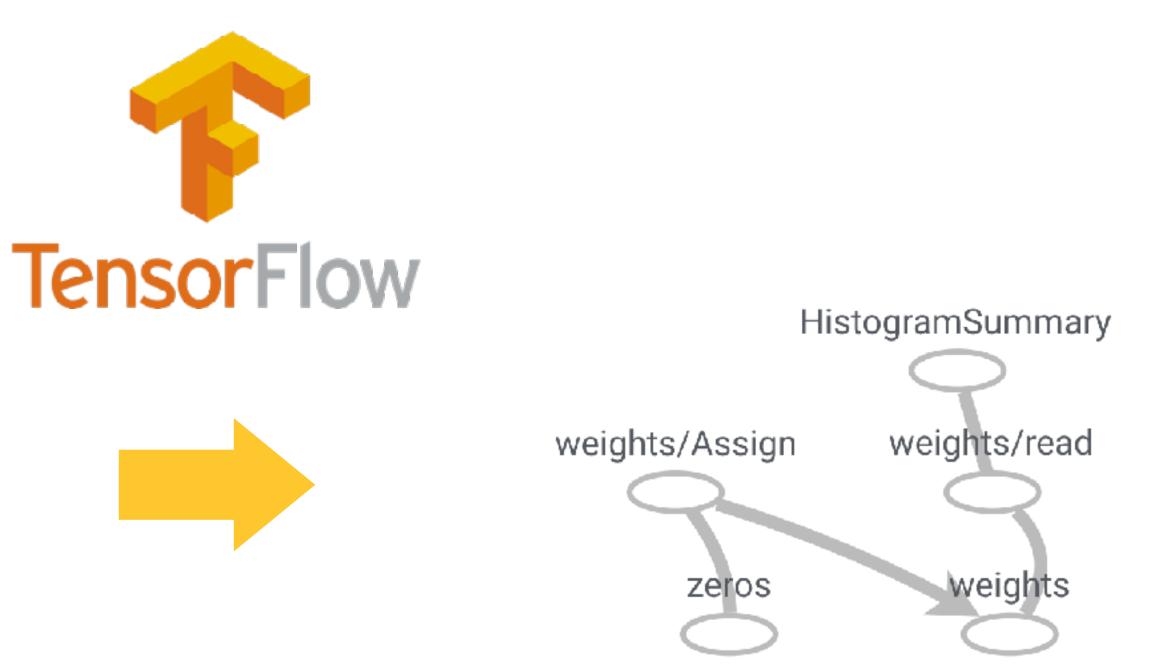




Low-level Dataflow Graph

```
zeros = tf.zeros([784, 10])
W = tf.Variable(zeros, name='weights')
tf.histogram_summary('weights', W)
```

High-level Program



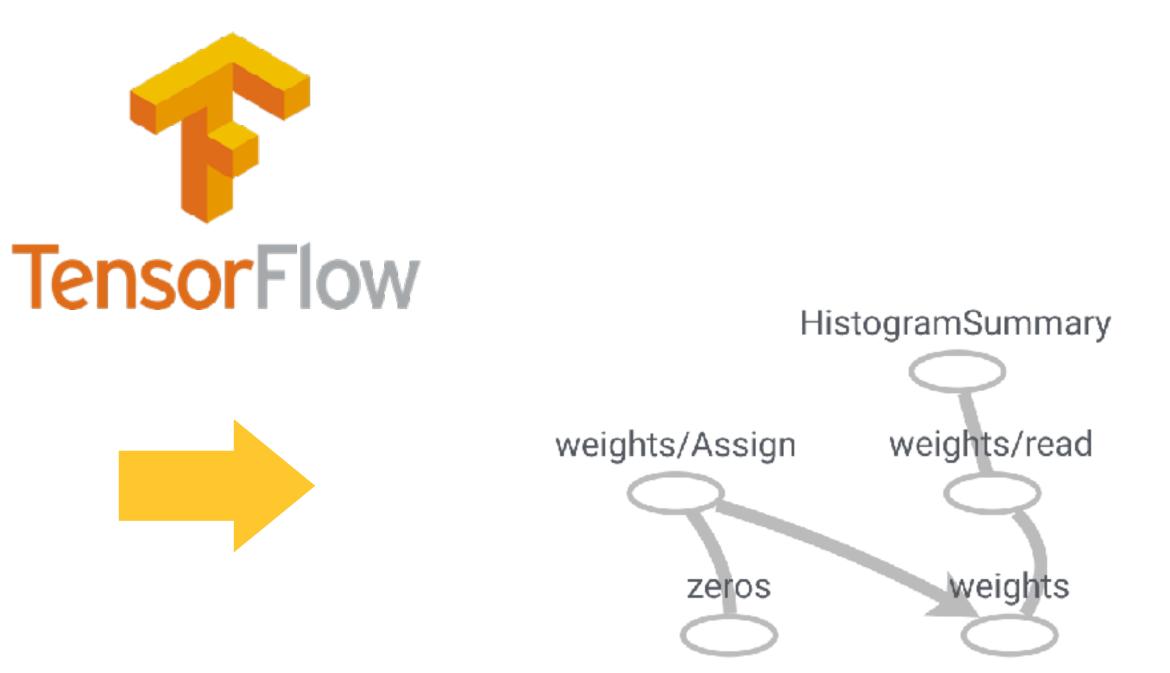
Low-level Dataflow Graph

Easy to implement models that supports distributed computation, various kinds of devices, and variety of learning algorithms.

```
zeros = tf.zeros([784, 10])
W = tf.Variable(zeros, name='weights')
tf.histogram_summary('weights', W)
```

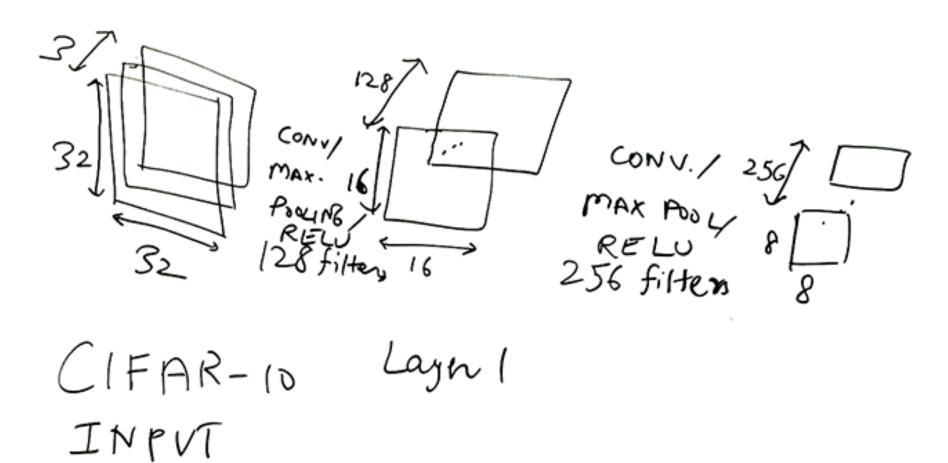
High-level Program

- Easy to implement models that supports distributed computation, various kinds of devices, and variety of learning algorithms.
 - Understanding the graph structure from the code can be challenging!



Low-level Dataflow Graph

Model developers use diagrams to understand and share *high-level* structures of their models.



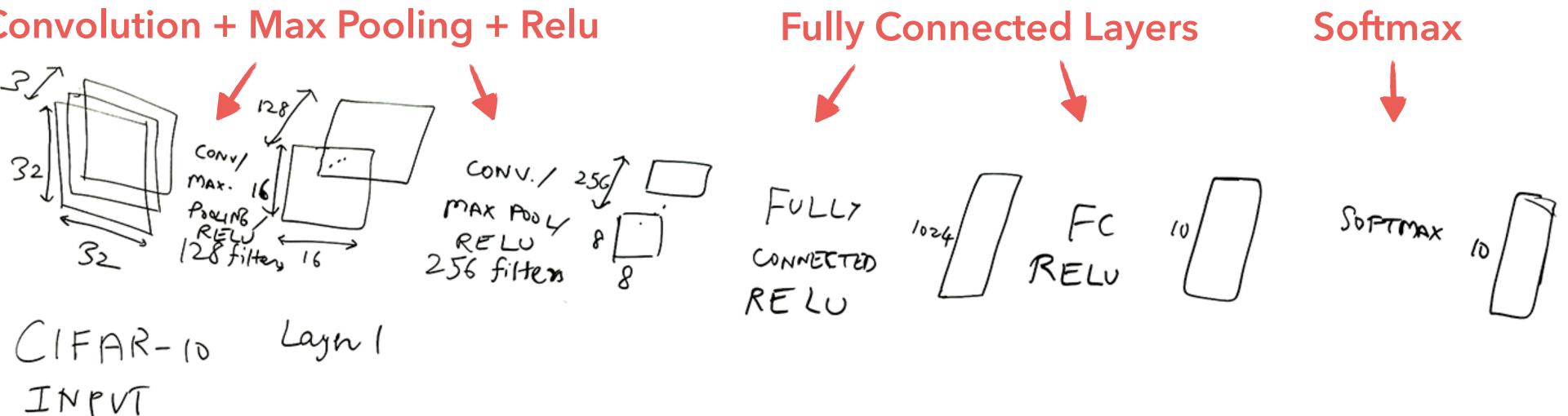
Typical hand-drawn diagram of a network by a researcher at Google

MAX POOL 250 FULLY 1024 FC 10 SOFTMAX 10 SOFTMAX 10 RELU 8 SOFTMAX 10 RELU 8 16 256 filters 8 051



Model developers use diagrams to understand and share high-level structures of their models.

Convolution + Max Pooling + Relu

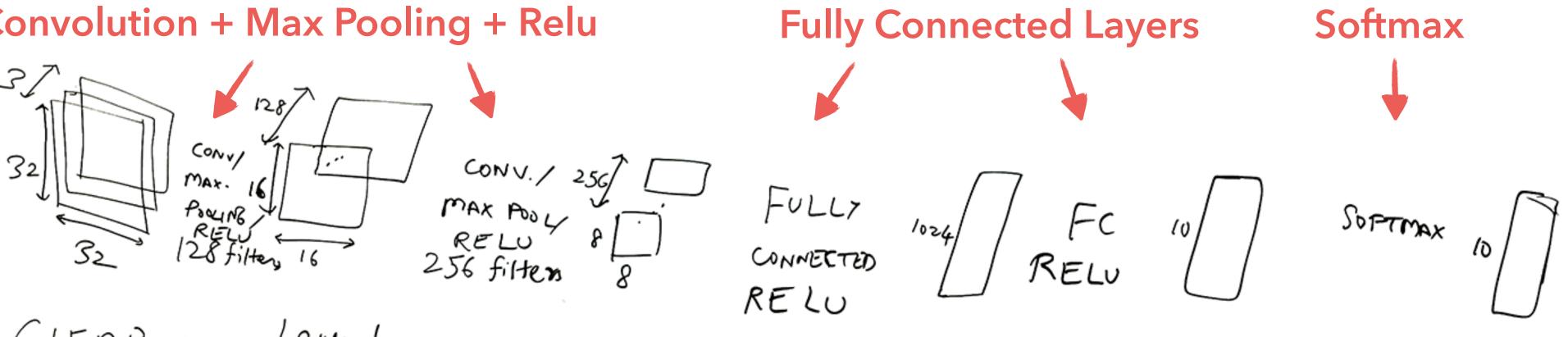


Typical hand-drawn diagram of a network by a researcher at Google



Model developers use diagrams to understand and share high-level structures of their models.

Convolution + Max Pooling + Relu



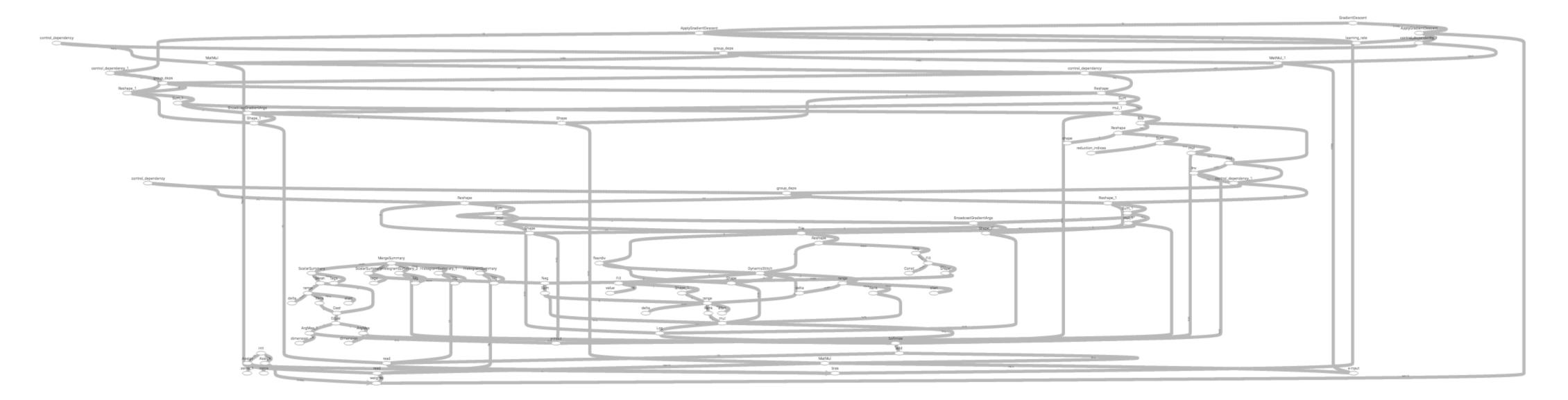
CIFAR-10 Lagn 1 INPVT

Typical hand-drawn diagram of a network by a researcher at Google

Automatic tool to visualize the model structure?



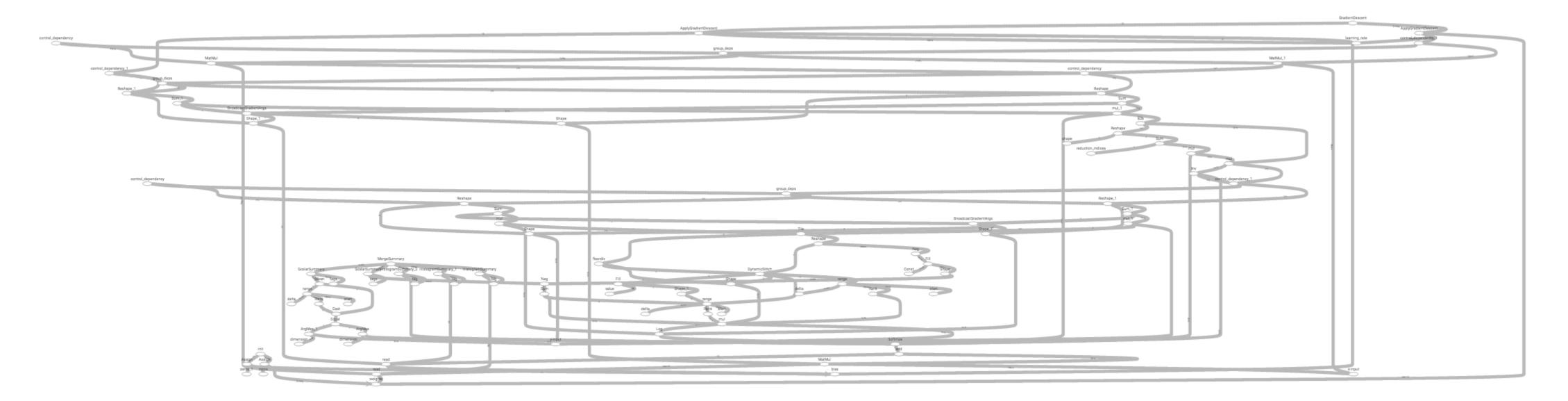
Standard graph drawing tools produce cluttered layouts



Sugiyama-style flow layout of a linear model (WX+B)



Standard graph drawing tools produce cluttered layouts

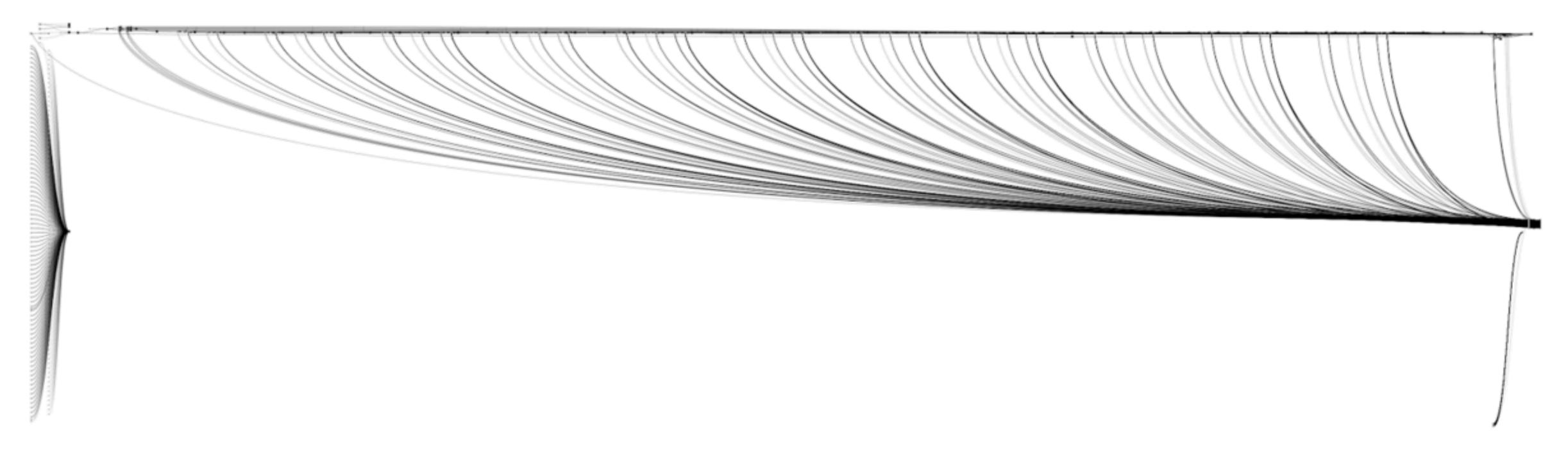


Sugiyama-style flow layout of a linear model (WX+B) TensorFlow's "Hello World"!





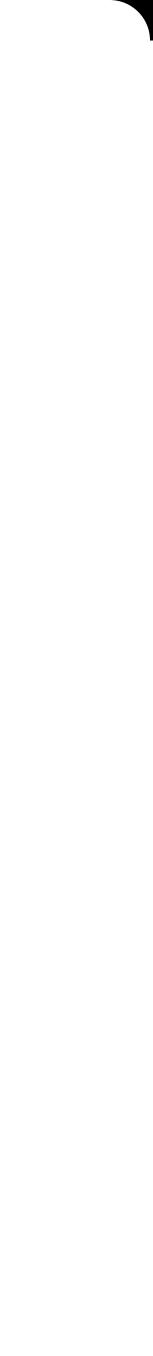
Standard graph drawing tools produce cluttered layouts

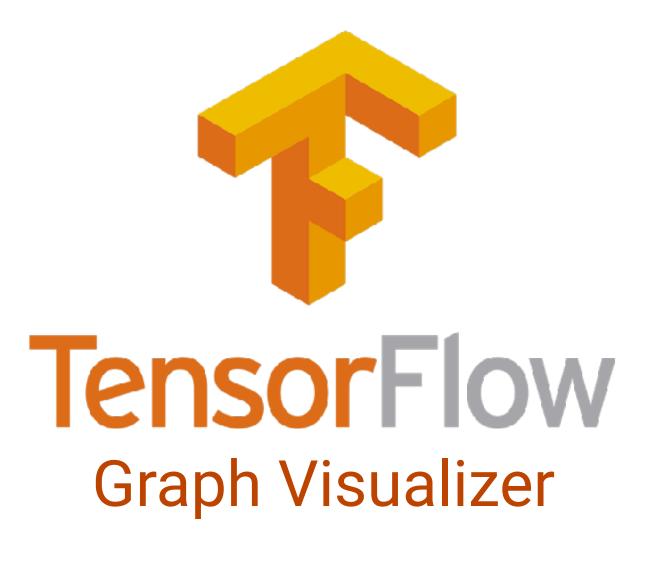


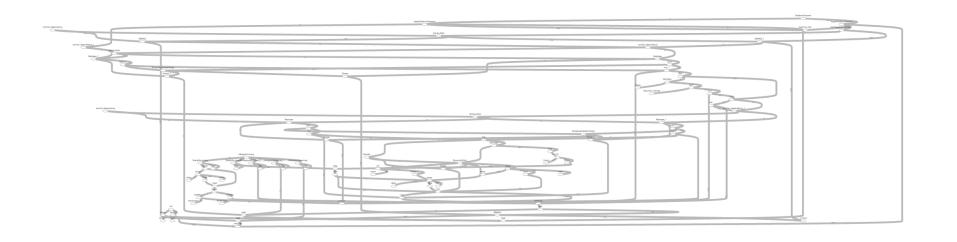
Sugiyama-style flow layout of a convolution network

TensorFlow **Graph Visualizer**

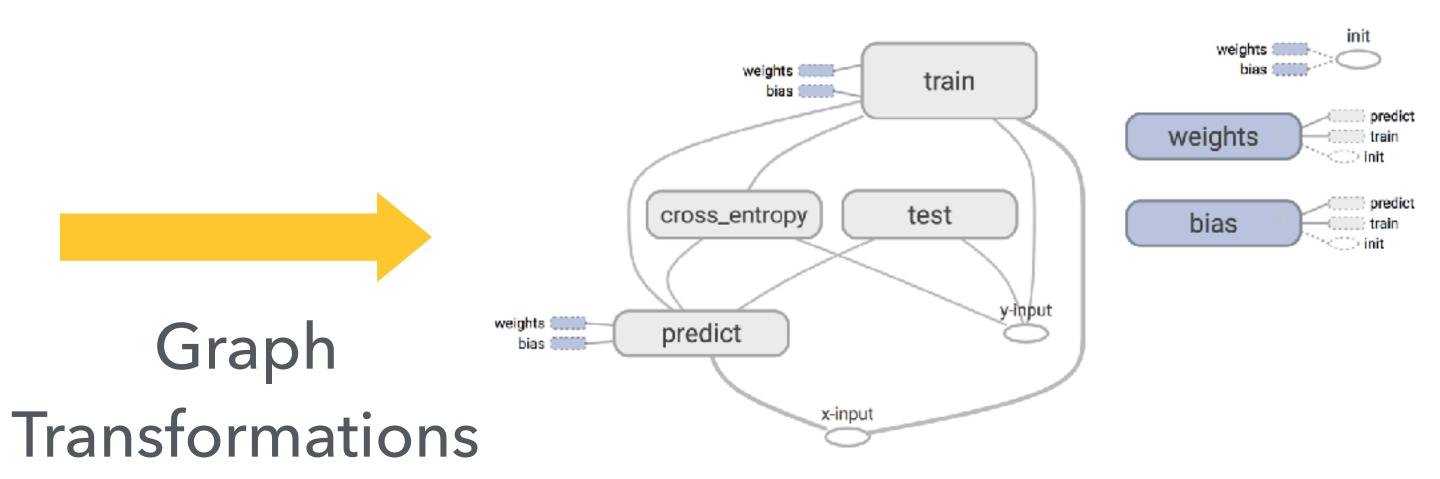
Help TensorFlow developers understand and inspect the structure of their models







Low-level Dataflow Graph



High-level Interactive Diagram

TensorFlow Graph Visualizer

Visualizing Dataflow Graphs of Deep Learning Models in TensorFlow

TensorFlow Graph Visualizer

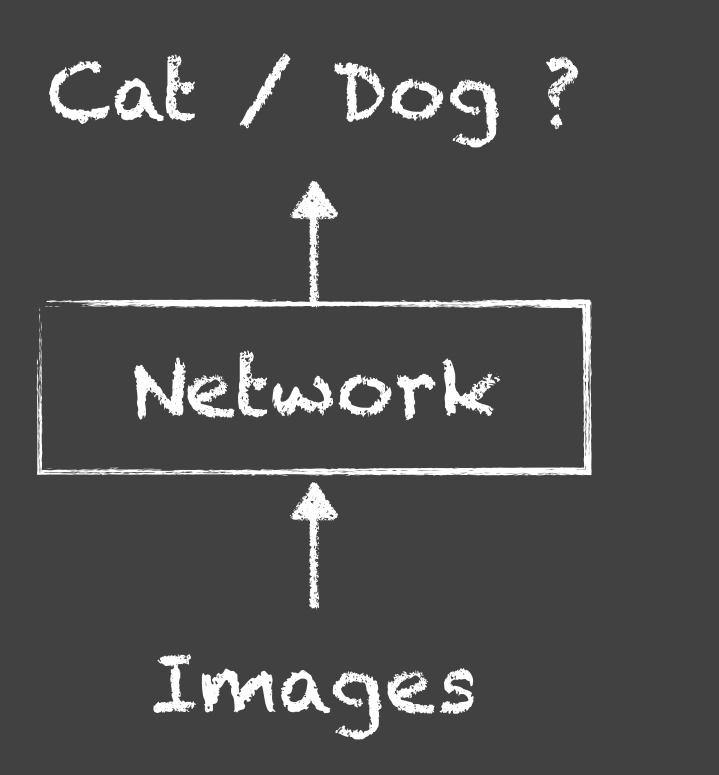
Visualizing Dataflow Graphs of Deep Learning Models in TensorFlow

Introduction

Explore a Convolutional Network

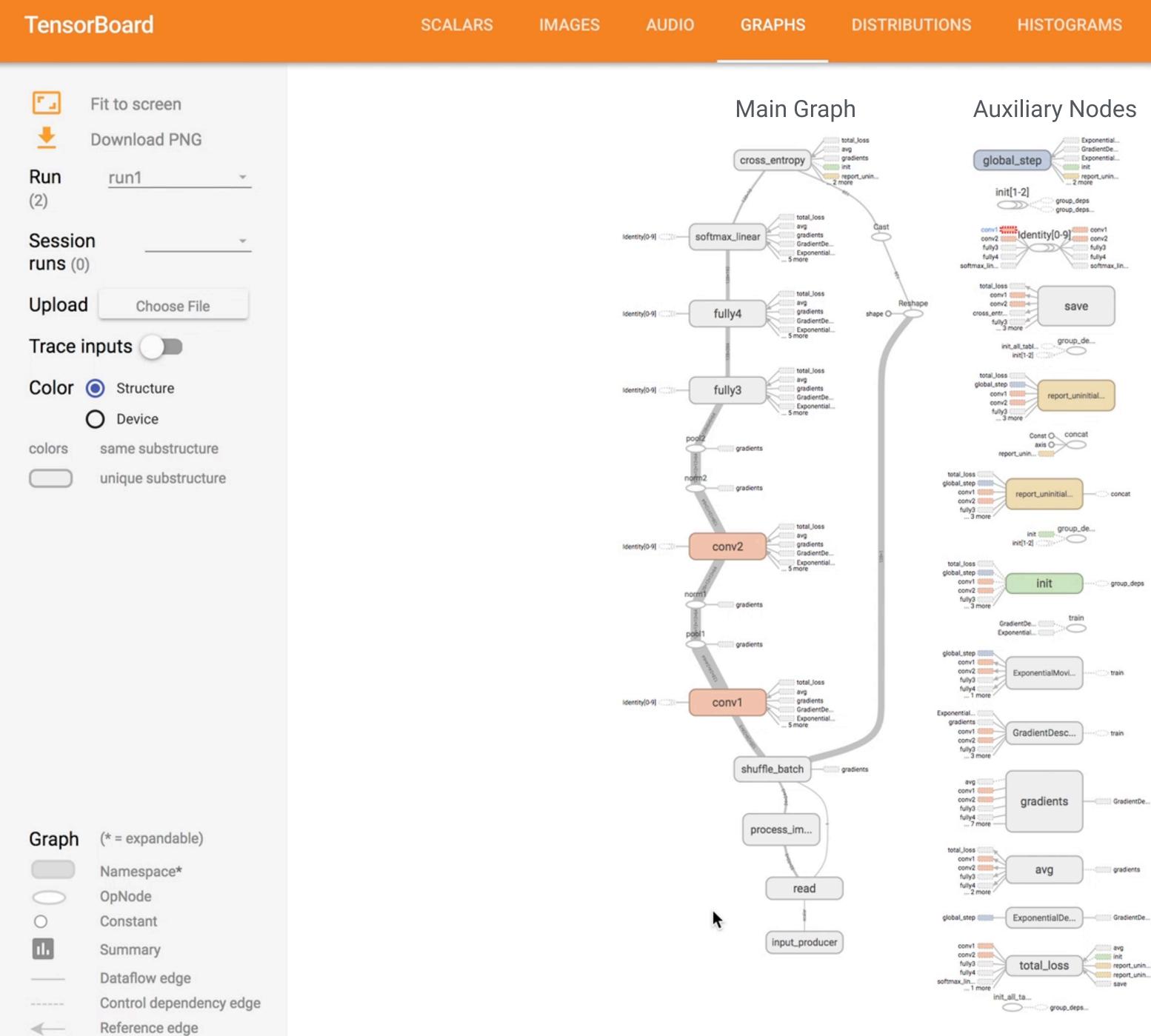
Transformation Strategies

Usage Pattern & Feedback



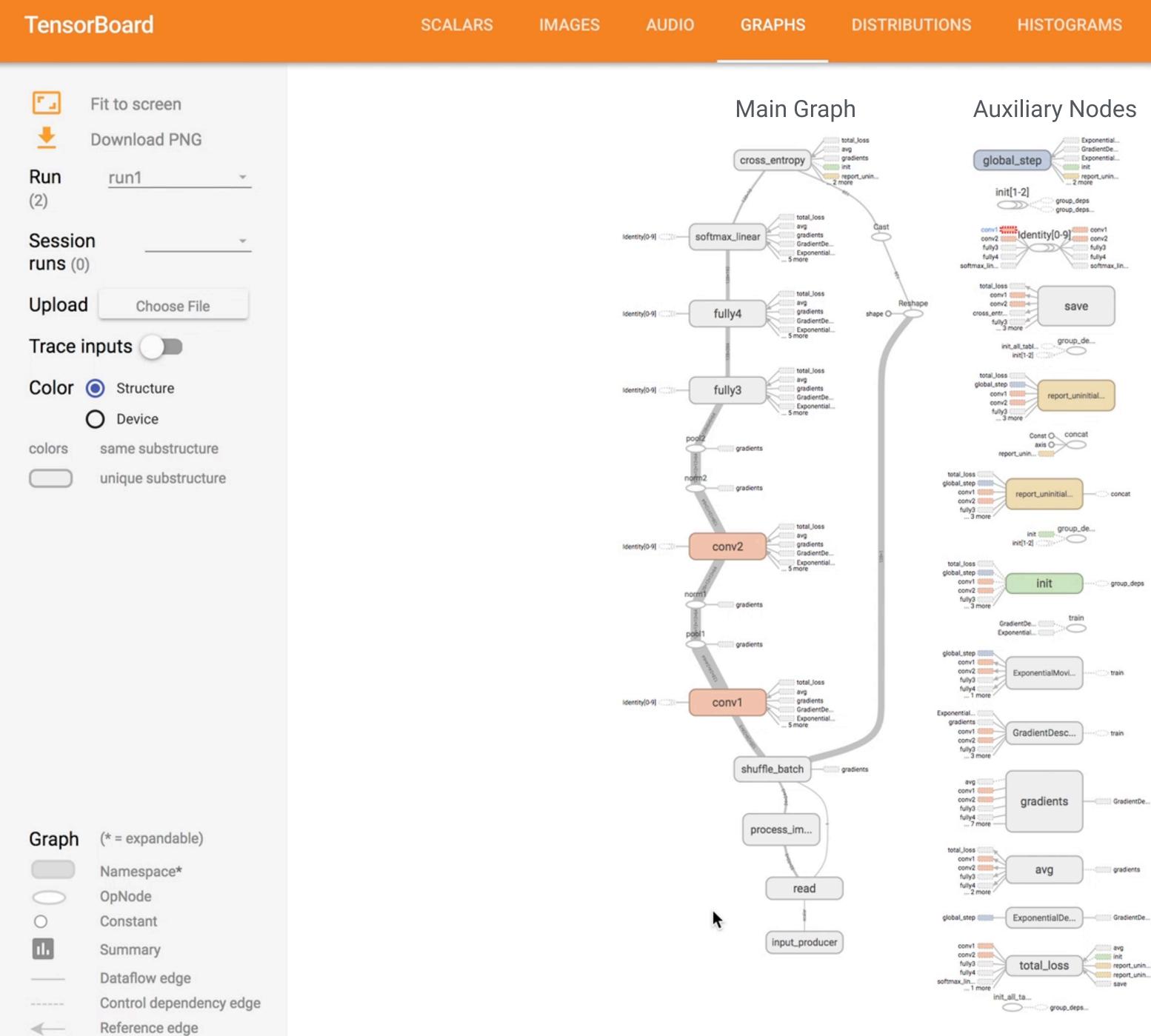
A dataflow for training a convolution network for image classification







-



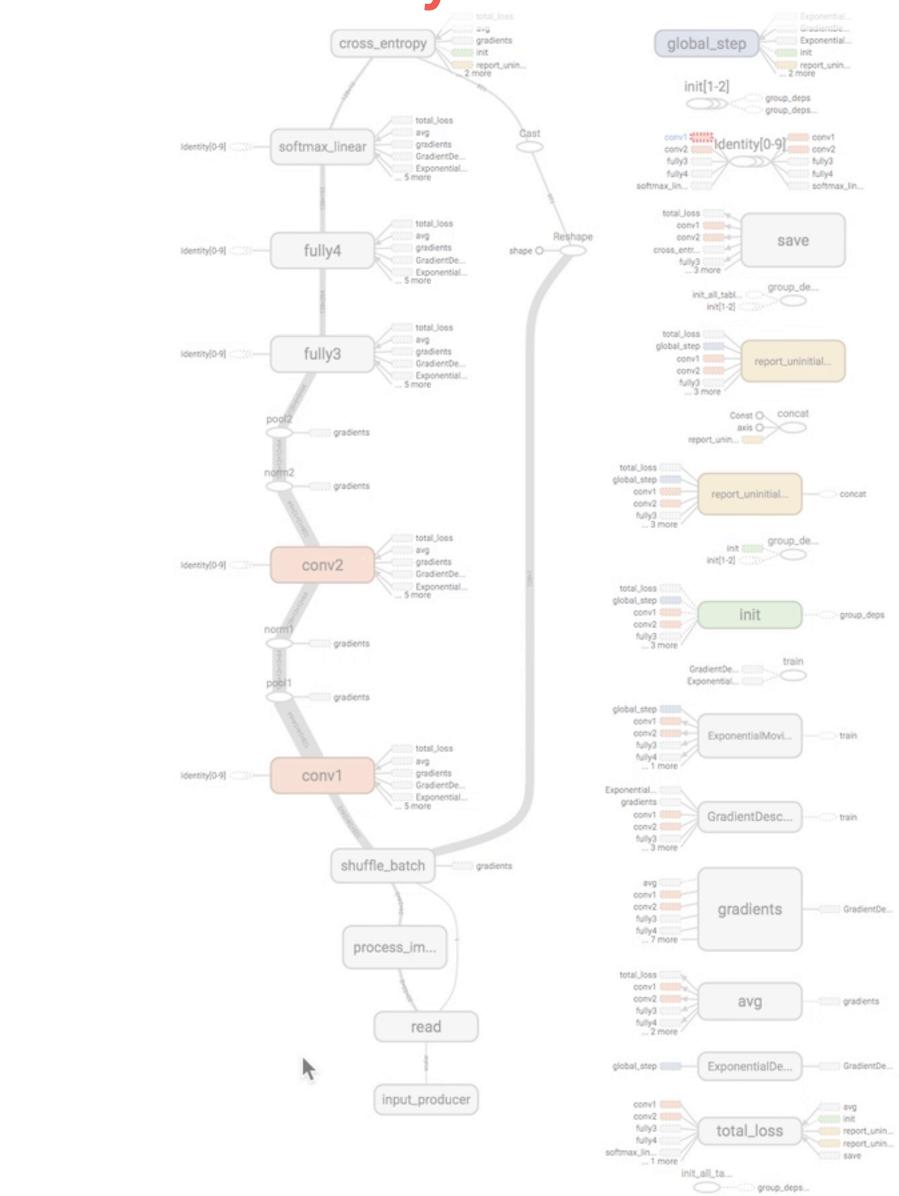


-

TensorBoard

<u>1</u>	Fit to screen	
<u>+</u>	Download PNG	
Run (2)	run1	~
Sessio runs (0)		Y
Upload	Choose File	
Trace i	nputs	
Color	Structure	
	O Device	
colors	same substructure	
	unique substructure	

TensorBoard has many kinds of visualizations



Graph (* = expandable) Namespace* OpNode 0 Constant 16

Summary Dataflow edge Control dependency edge Reference edge

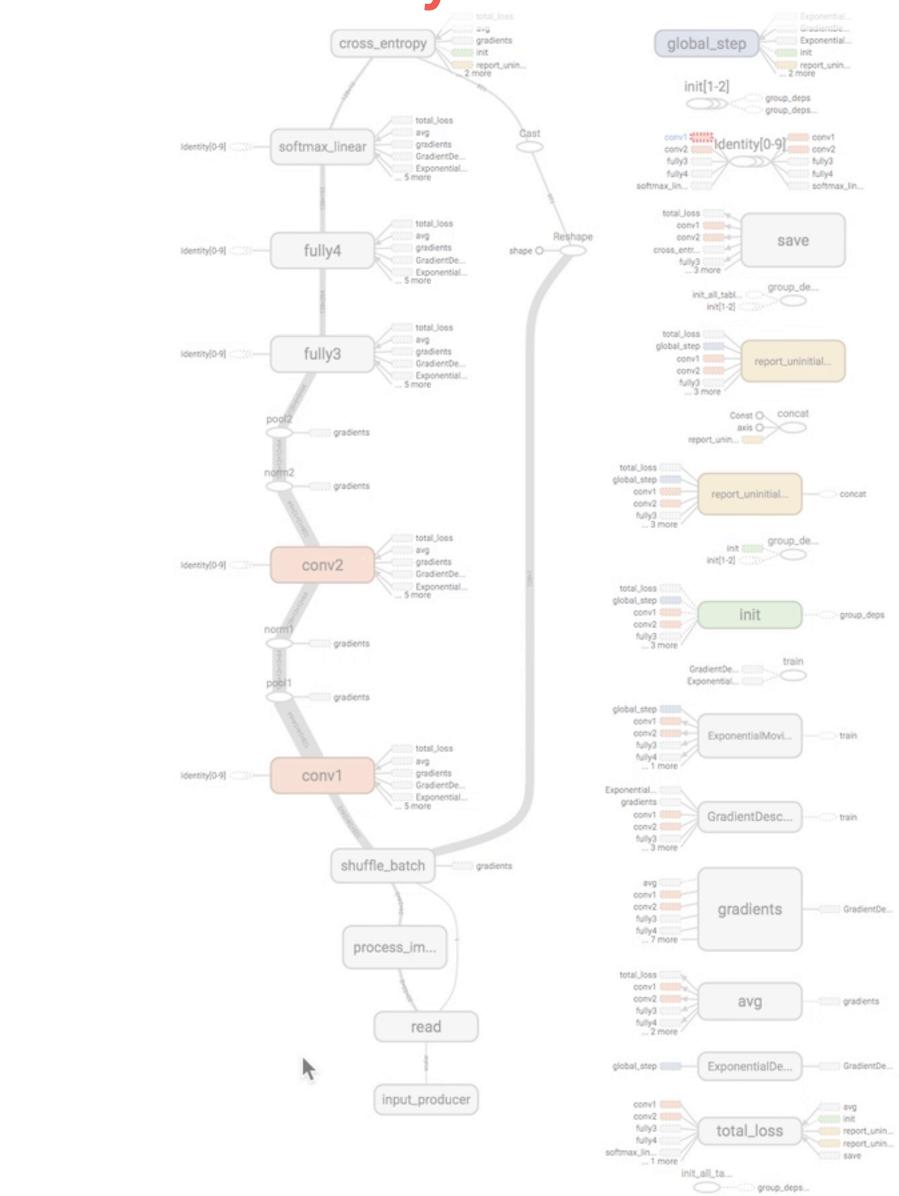


-

TensorBoard

<u>1</u>	Fit to screen	
<u>+</u>	Download PNG	
Run (2)	run1	~
Sessio runs (0)		Y
Upload	Choose File	
Trace i	nputs	
Color	Structure	
	O Device	
colors	same substructure	
	unique substructure	

TensorBoard has many kinds of visualizations



Graph (* = expandable) Namespace* OpNode 0 Constant 16

Summary Dataflow edge Control dependency edge Reference edge



-

		_		
Top	OOF		0.2	
Ten	SOF	D10	7.11	
	~~~	_		

F ...

ᆂ

Run

Session

runs (0)

Upload

colors

Graph

0

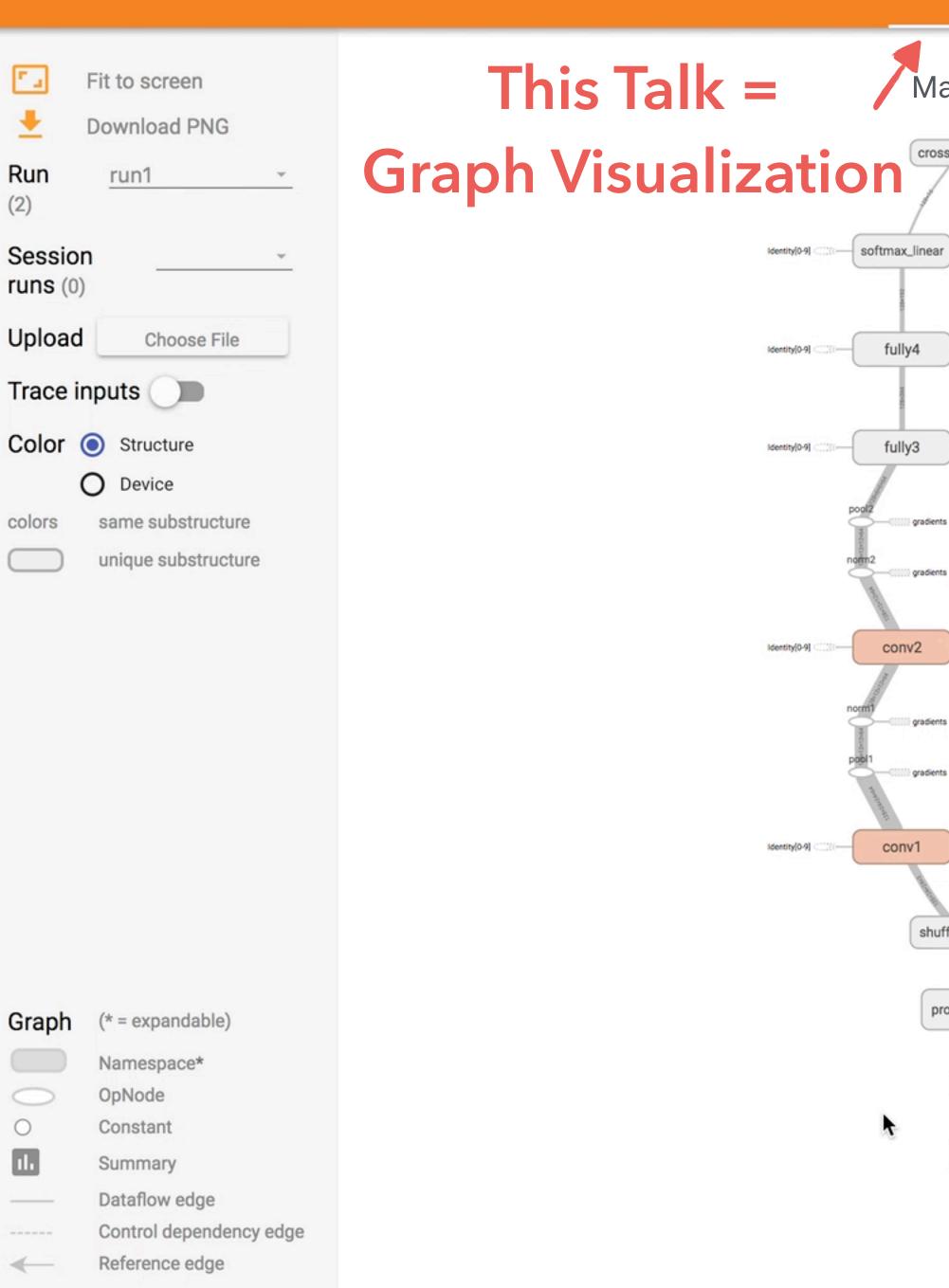
ıl.

_____

----

----

(2)





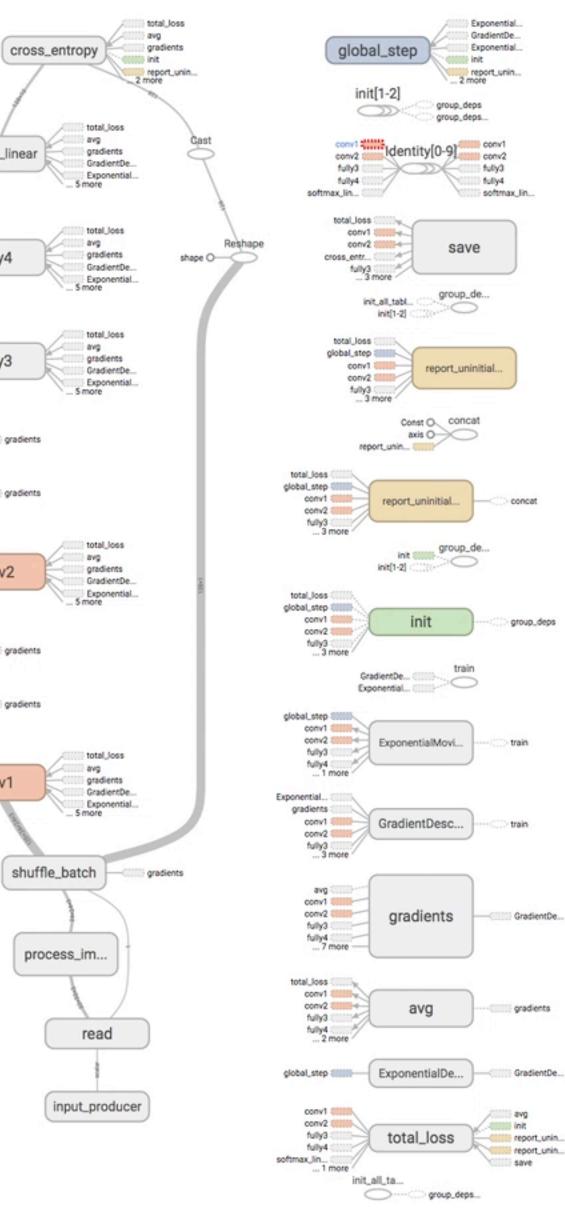
gradients

oradients

gradients

gradients

## Auxiliary Nodes





-

		_		
Top	OOF		0.2	
Ten	SOF	<b>D10</b>	7.11	
	~~~	_		

F ...

ᆂ

Run

Session

runs (0)

Upload

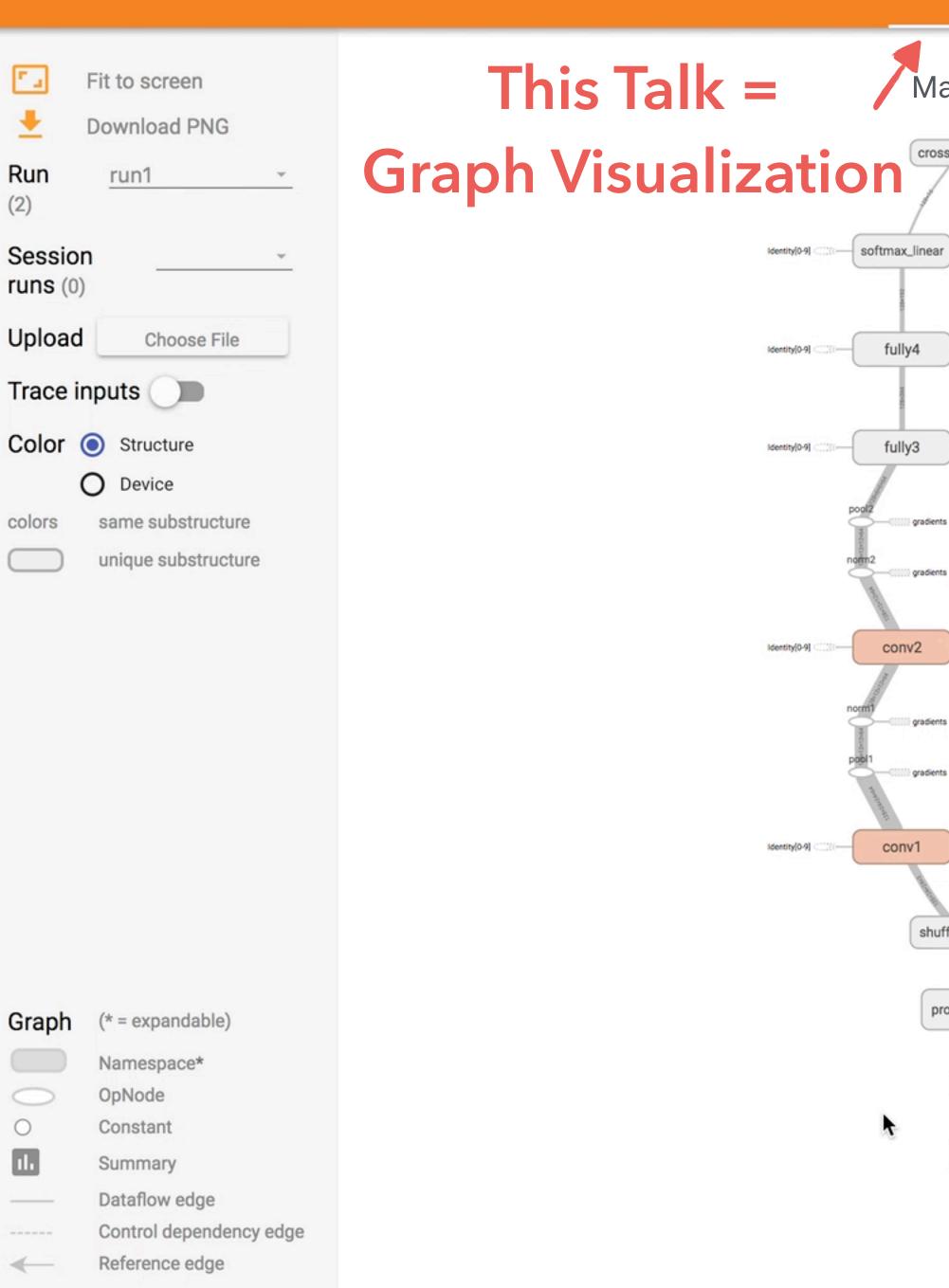
colors

Graph

0

ıl.

(2)





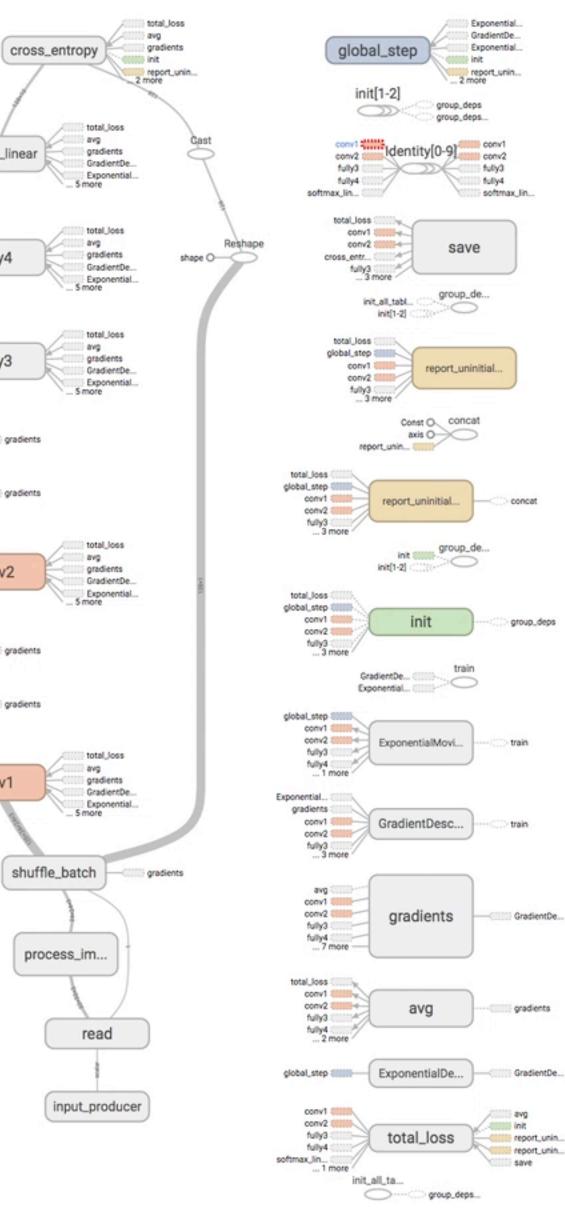
gradients

oradients

gradients

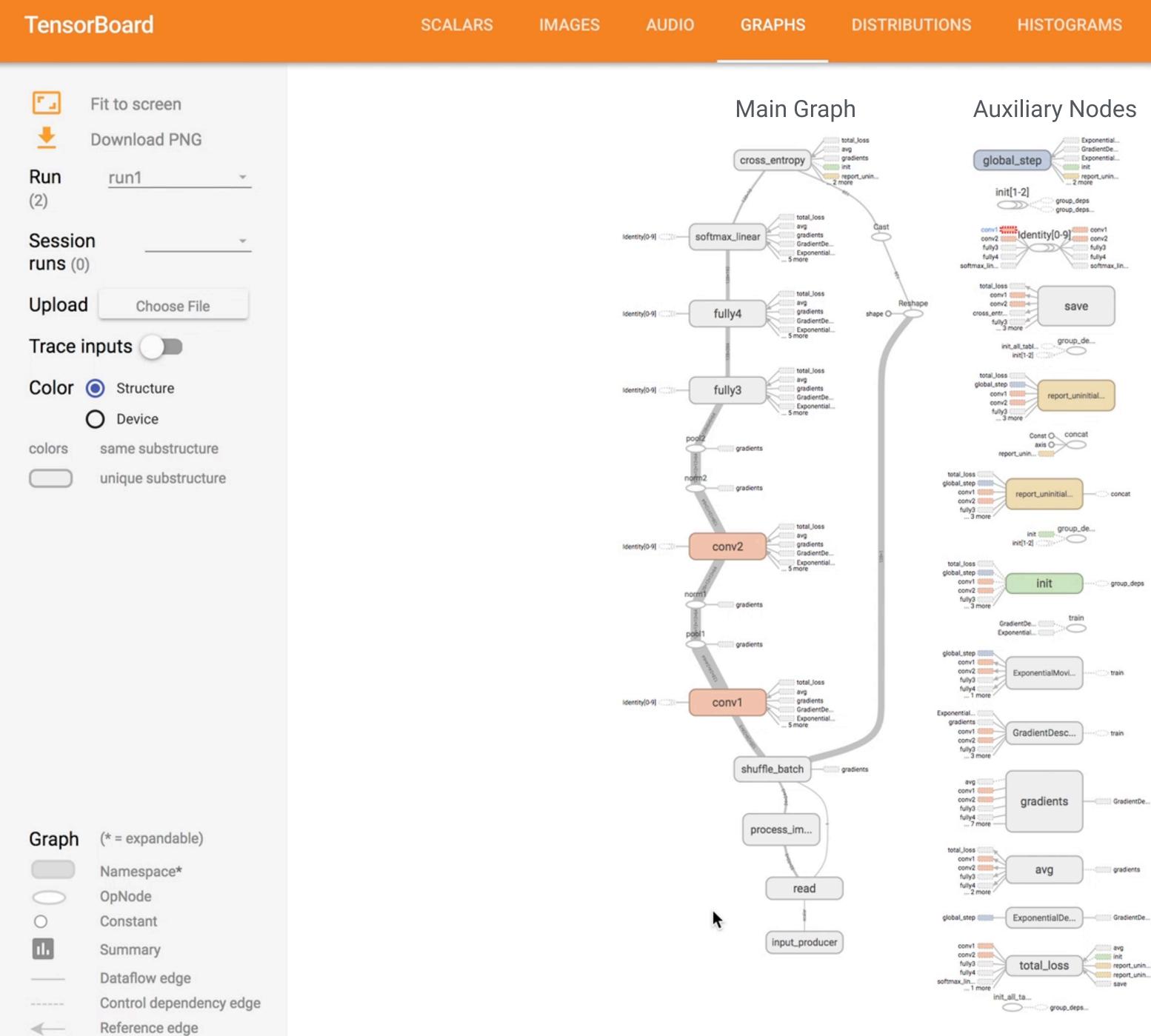
gradients

Auxiliary Nodes



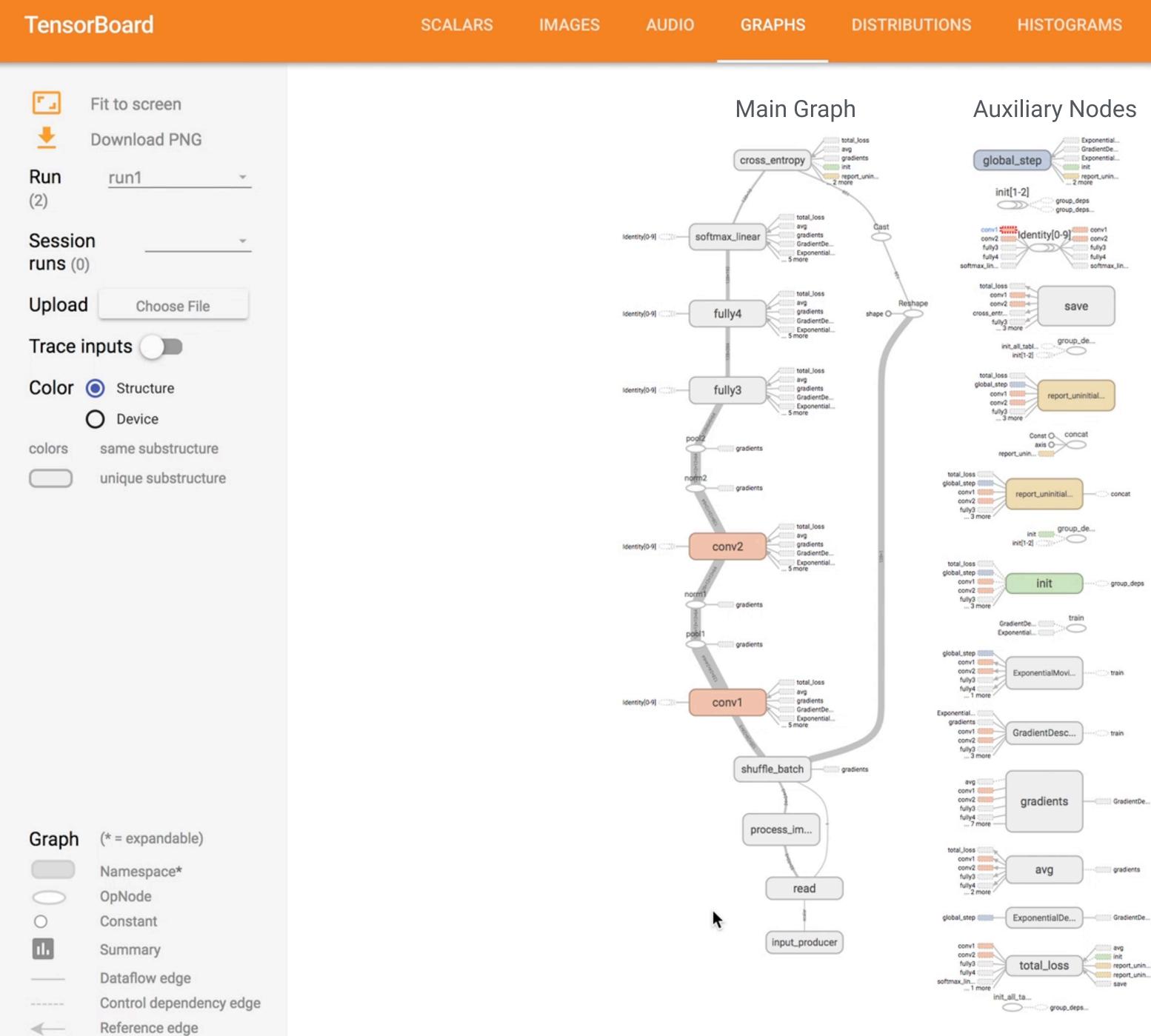


-





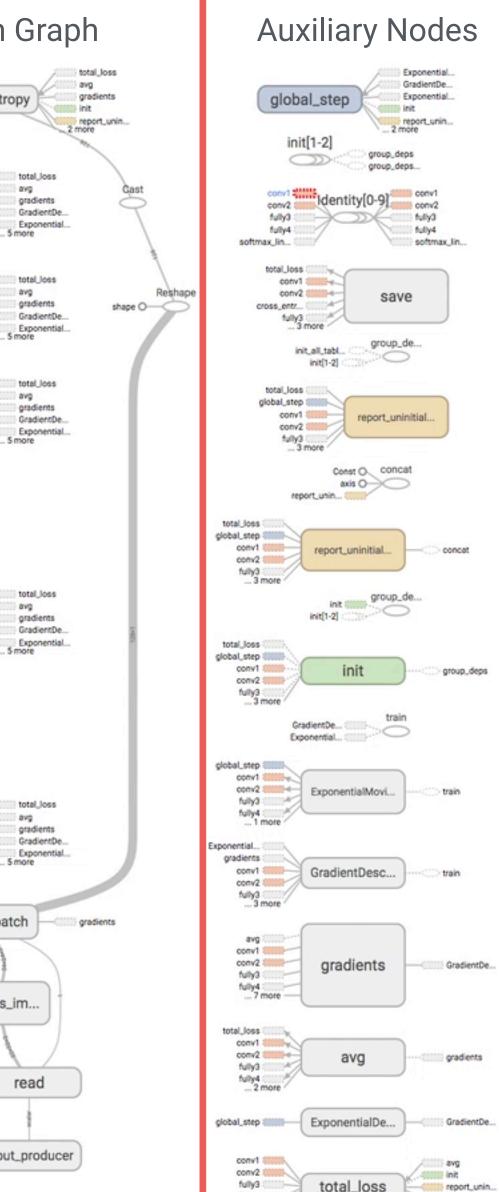
-





-

TensorBoa	rd	SCALARS	IMAGES	AUDIO	GRAPHS
Run run (2) Session runs (0) Upload Trace inputs Color O Stru O Dev colors same	choose File Choose File ucture substructure substructure	lain Gra oputation f	-	Identity[0-9] [U Identity[0-9] [U Identi	Main G cross_entropy ax_linear ax_linear ax_linear by cross_entropy ax_linear by cross_entropy cross_
 Name OpNo Const Summ Datafl Control 	ant			Þ	process_im.



Auxiliary Nodes

Exponential GradientDe

softmax_lin_

> group_deps

---- train

train

GradientDe ...

----- gradients

avg.

Cititis report_unin_

save

cititi report_unin__

-country init

- cossi init report_unin_

group_deps group_deps...

save

report_uninitial...

init group_de...

init

GradientDesc...

gradients

total_loss

init_all_ta... group_deps...

fully4

softmax_lin____1 more

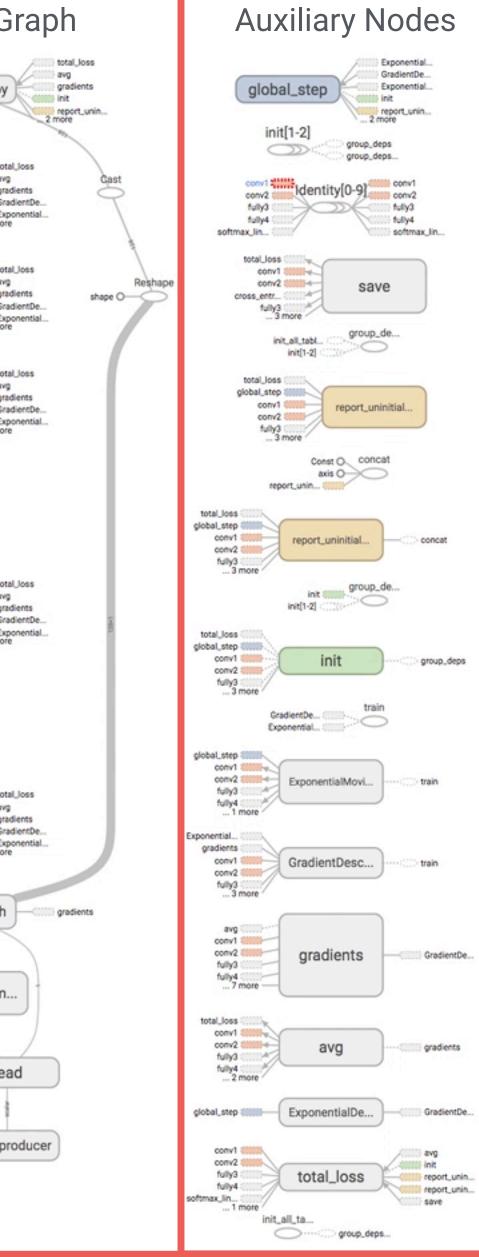






Tenso	orBoard	SCALARS	IMAGES	AUDIO	GRAPHS
Colors	 Choose File Choose File Choose File Structure Structure Device same substructure unique substructure 	Main Gra computation f	-	Identity[0-9] filentity[0-9] filenti	dents gradients gradients gradients gradients gradients gradients gradients gradients gradients gradients gradients gradients gradients gradients gradients gradients gradients
Graph	(* = expandable) Namespace* OpNode Constant Summary Dataflow edge Control dependency edge			k	process_im.
-	Reference edge				



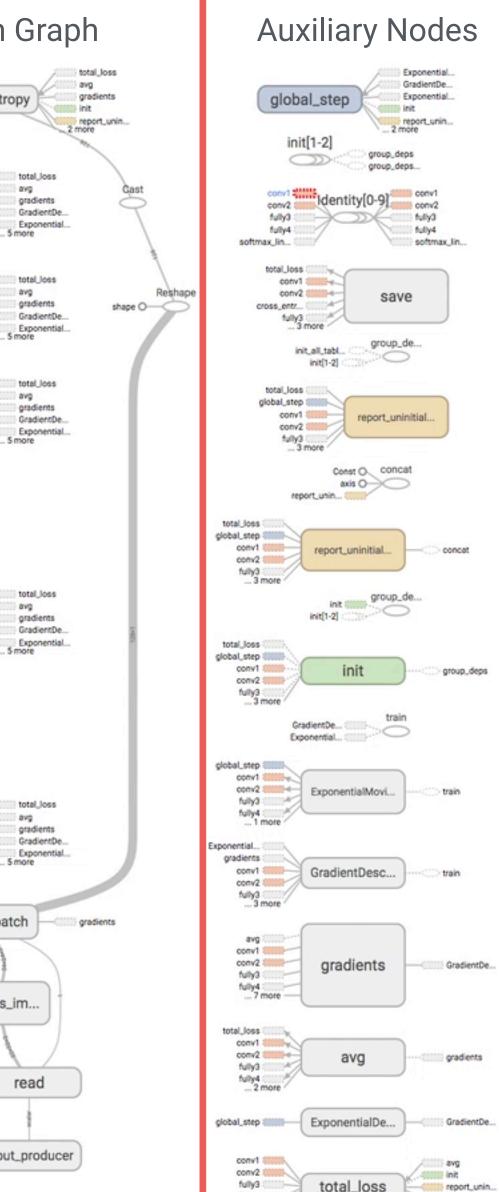


Auxiliary Nodes

less important operations that are extracted from the main graph



TensorBoa	rd	SCALARS	IMAGES	AUDIO	GRAPHS
Run run (2) Session runs (0) Upload Trace inputs Color O Stru O Dev colors same	choose File Choose File ucture substructure substructure	lain Gra oputation f	-	Identity[0-9] [U Identity[0-9] [U Identi	Main G cross_entropy ax_linear ax_linear ax_linear by cross_entropy ax_linear by cross_entropy cross_
 Name OpNo Const Summ Datafl Control 	ant			Þ	process_im.



Auxiliary Nodes

Exponential GradientDe

softmax_lin_

> group_deps

---- train

train

GradientDe ...

----- gradients

avg.

Cititis report_unin_

save

cititi report_unin__

-country init

- cossi init report_unin_

group_deps group_deps...

save

report_uninitial...

init group_de...

init

GradientDesc...

gradients

total_loss

init_all_ta... group_deps...

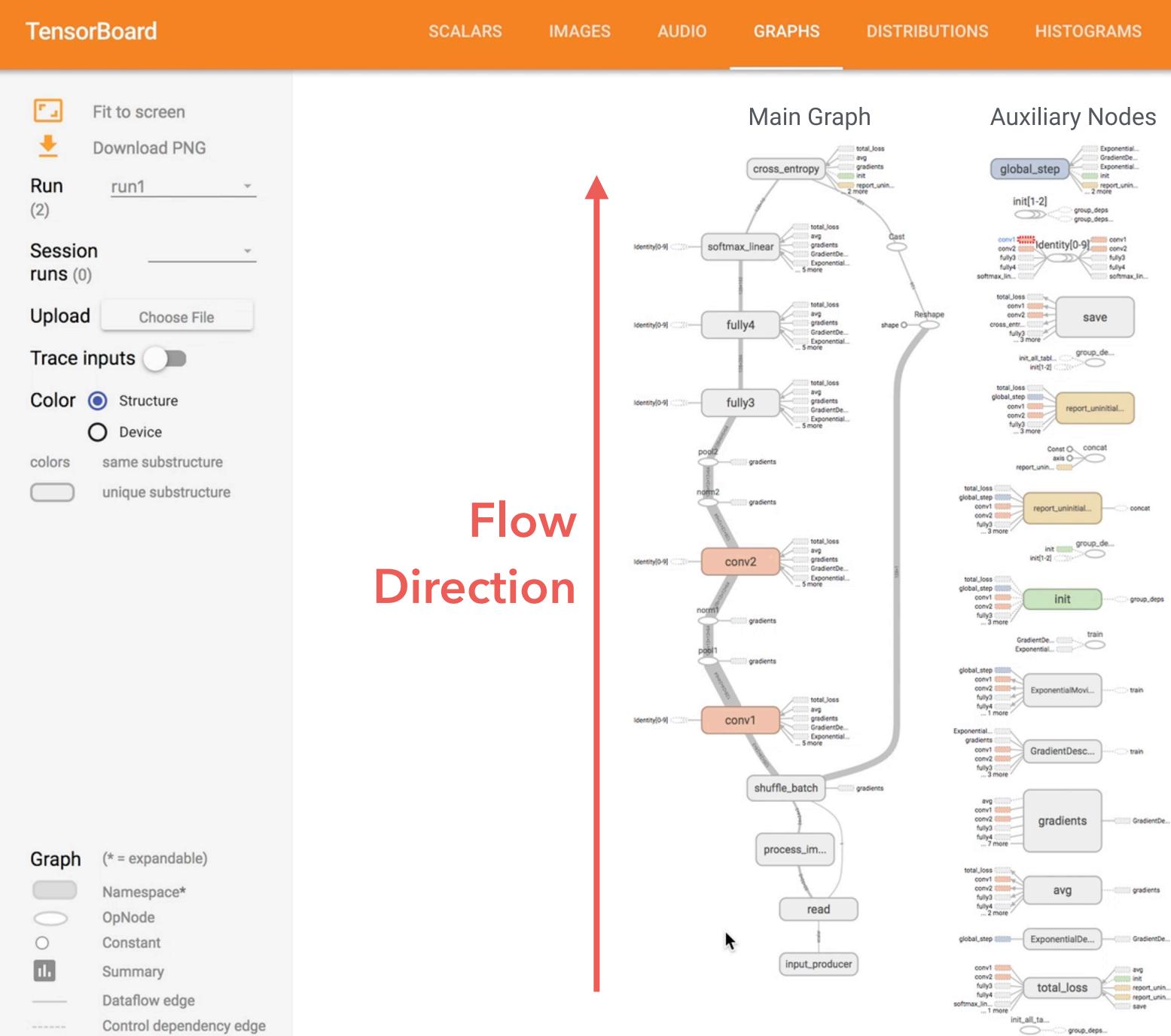
fully4

softmax_lin____1 more



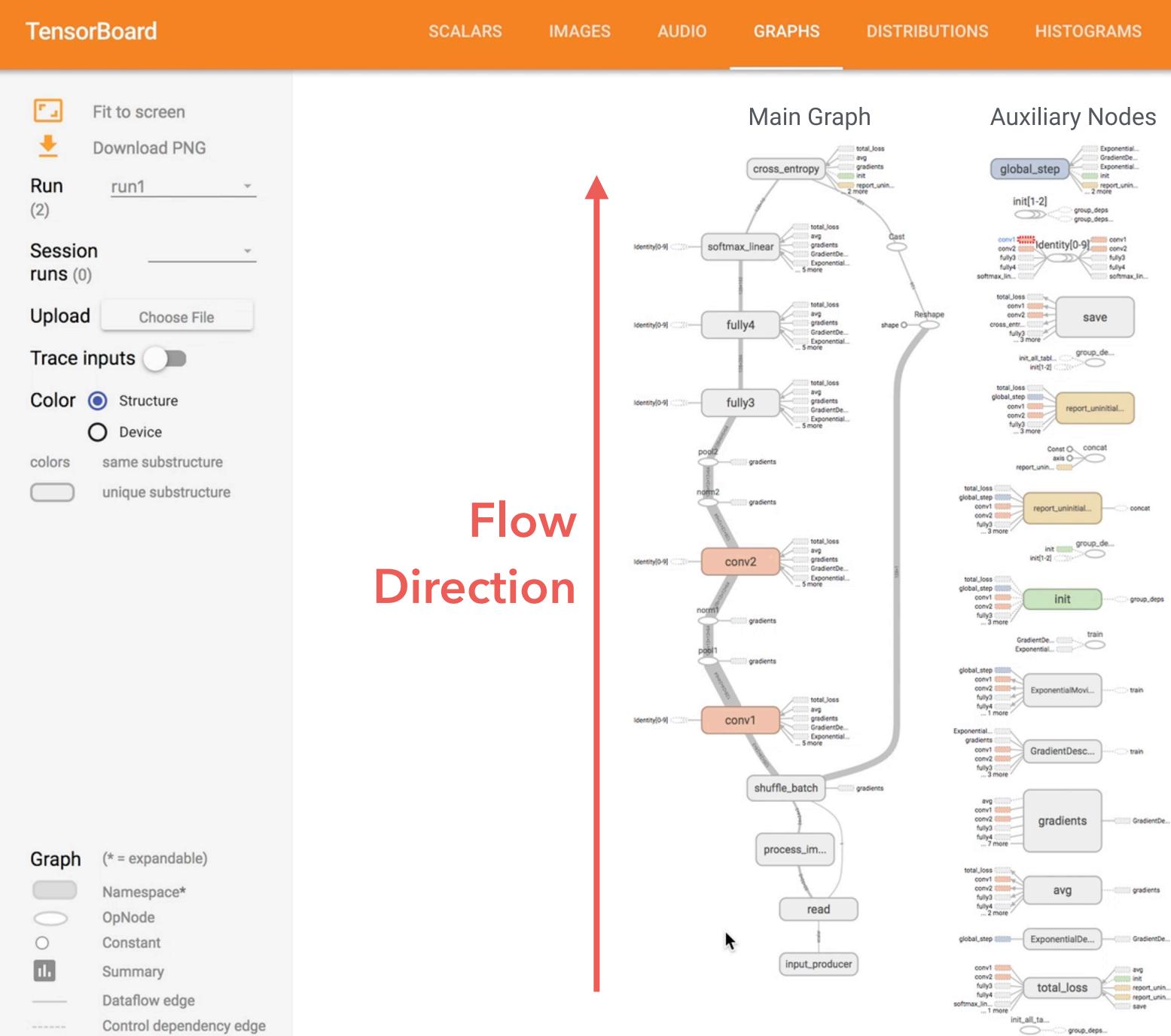






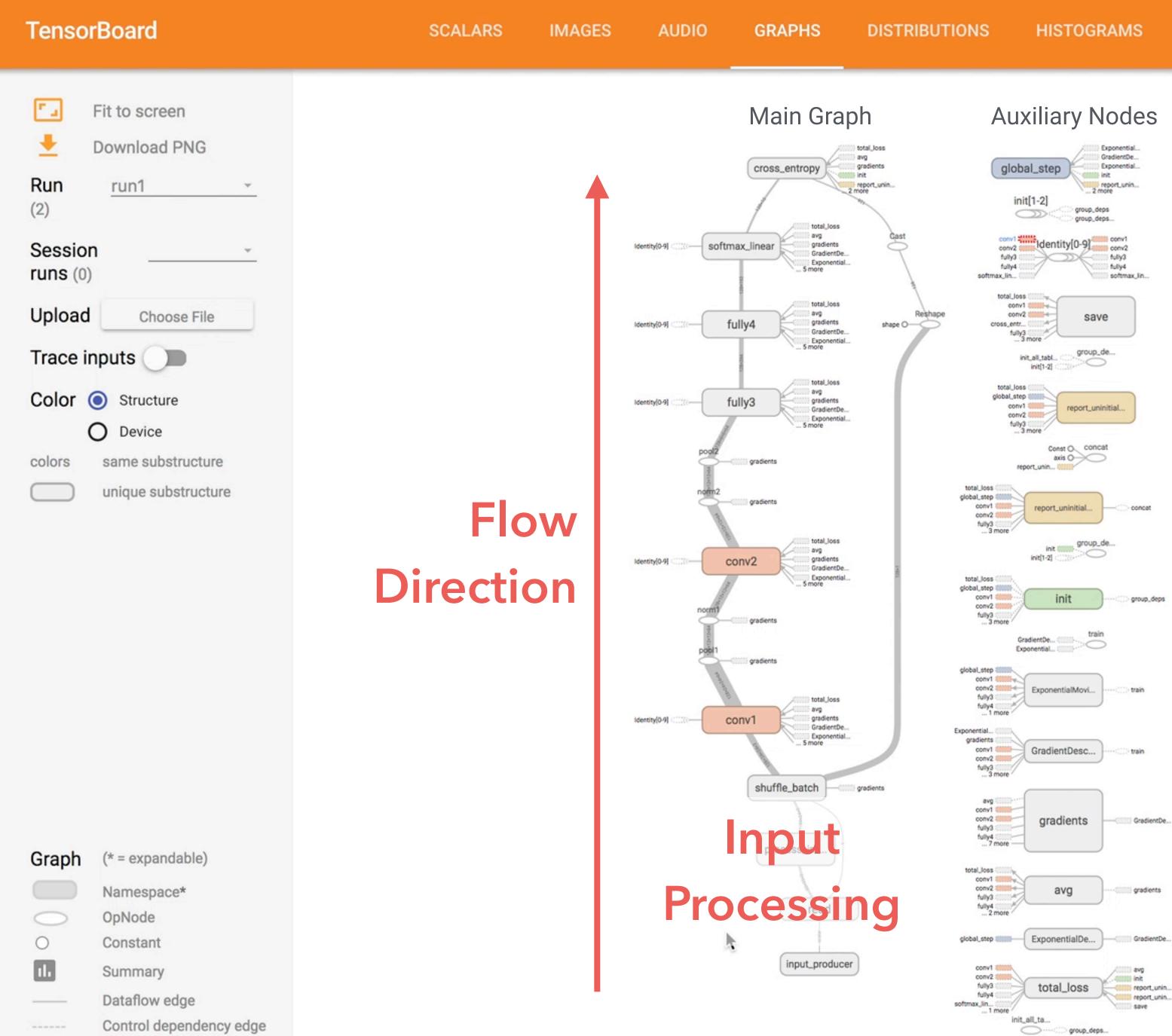


-





-



-

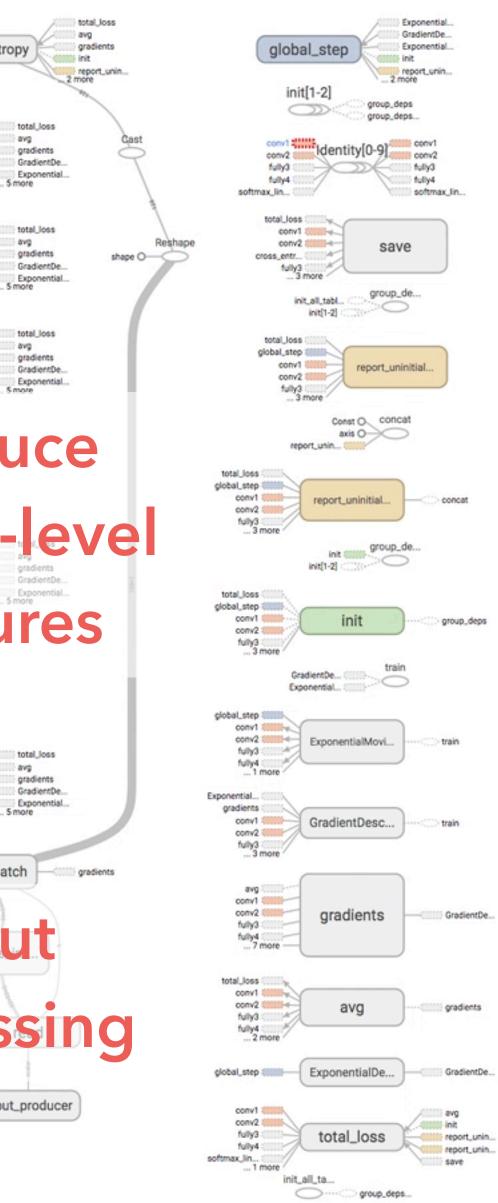


-

Tens	orBoard	SCALARS	IMAGES	AUDIO	GRAPHS
Colors Graph	(0) ad Choose File inputs Structure Device same substructure unique substructure (* = expandable) Namespace* OpNode Constant Summary Dataflow edge			Identity[0-9] CONTRACTOR OF THE Identity[0-9] CONTRACTOR OF THE IDENTITY [0-9] CONTRACTOR OF THE ID	gradients
	Control dependency edge				



Auxiliary Nodes



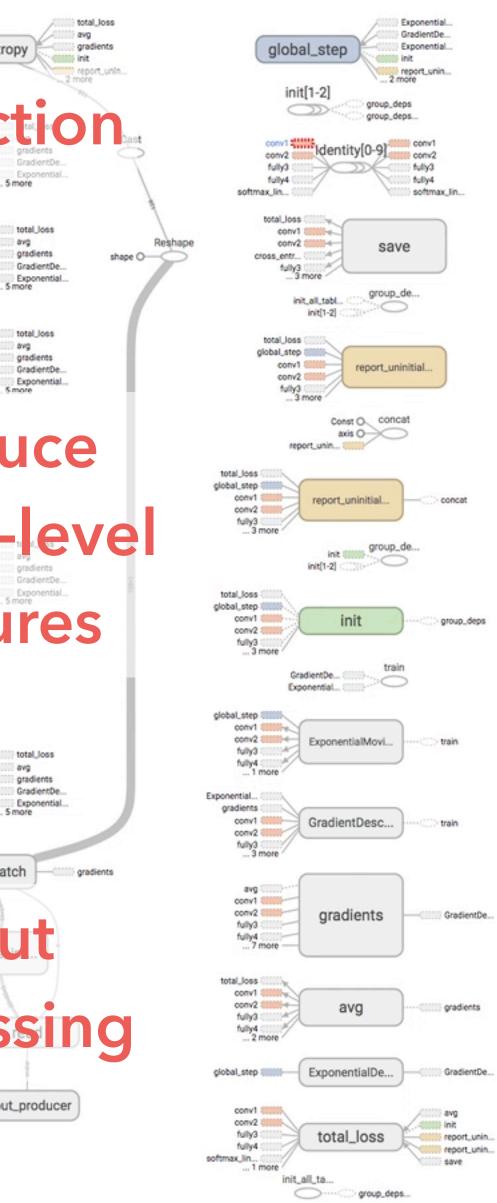


\$

Image: Session runs (n) Upload choose File Trace inputs Image:	TensorBoard	SCALARS	IMAGES	AUDIO	GRAPHS
Dataflow edge	Download PNG Run run1 (2) Session runs (0) Upload Choose File Trace inputs Image: Ima			Identity[0-9] COLOURS of from Colors of the software of the so	cross_entrop
	 Namespace* OpNode Constant Summary 			Identity[0-9]	onv1

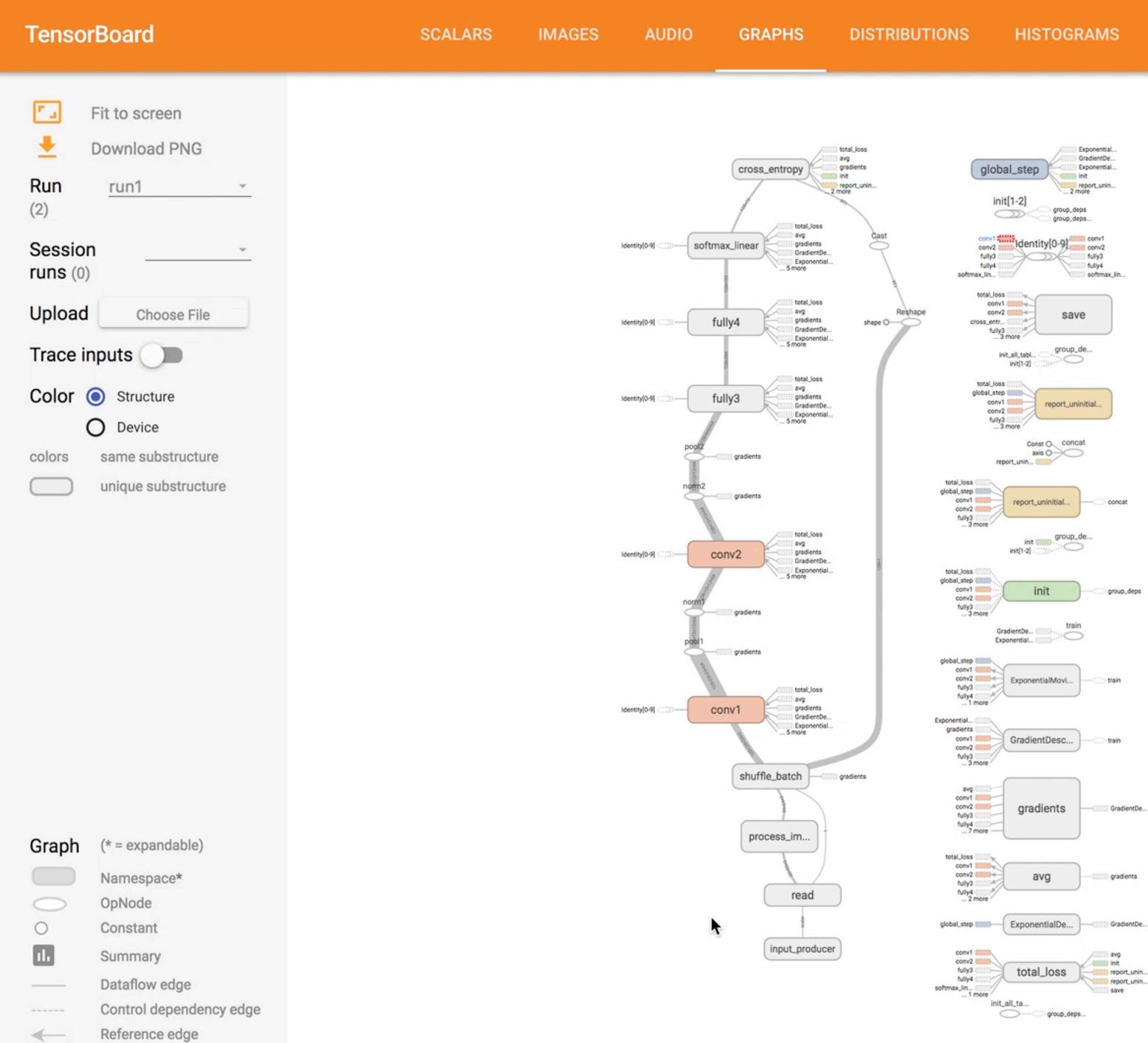


Auxiliary Nodes





\$

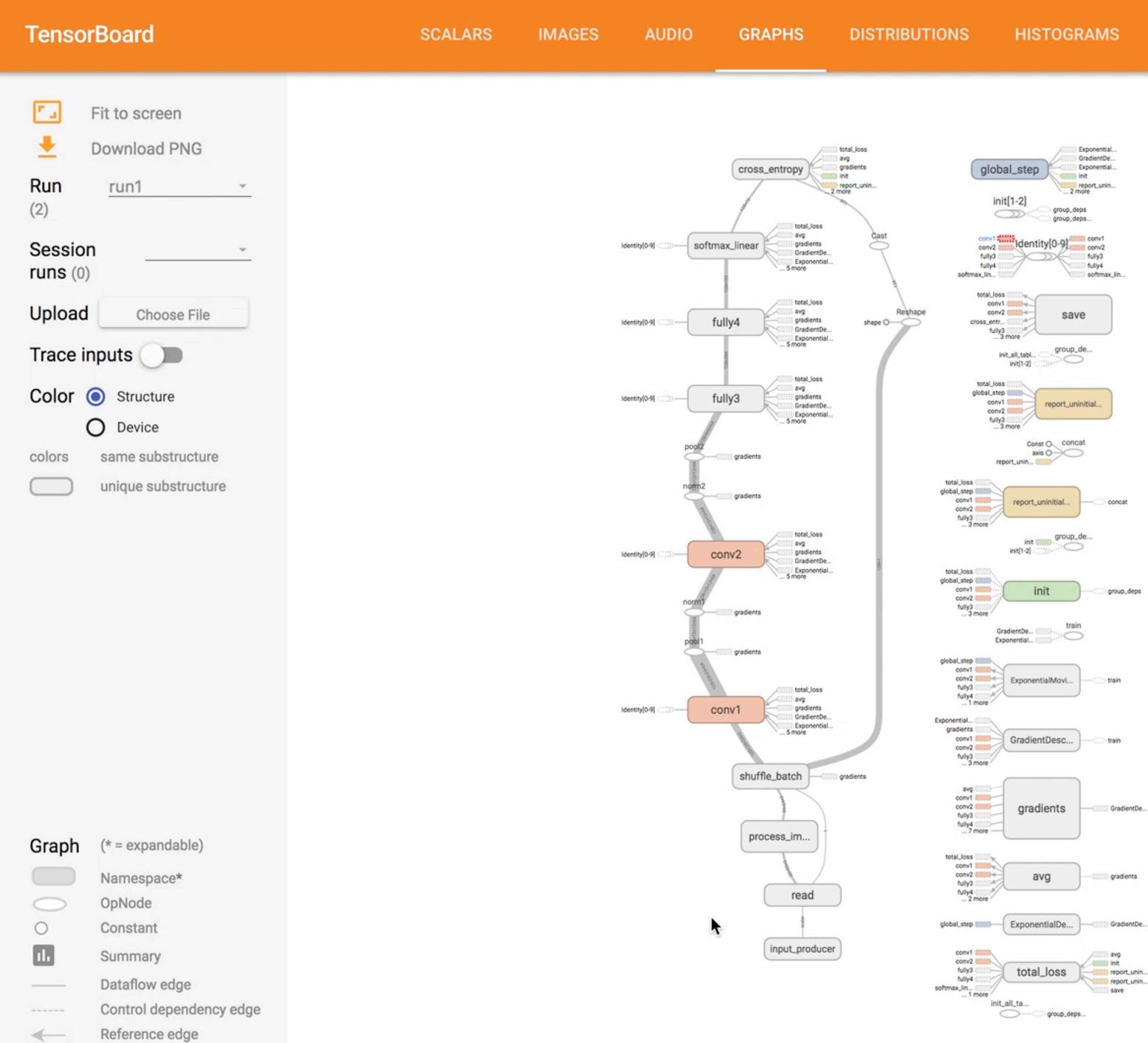


 \leftarrow



-

?



 \leftarrow

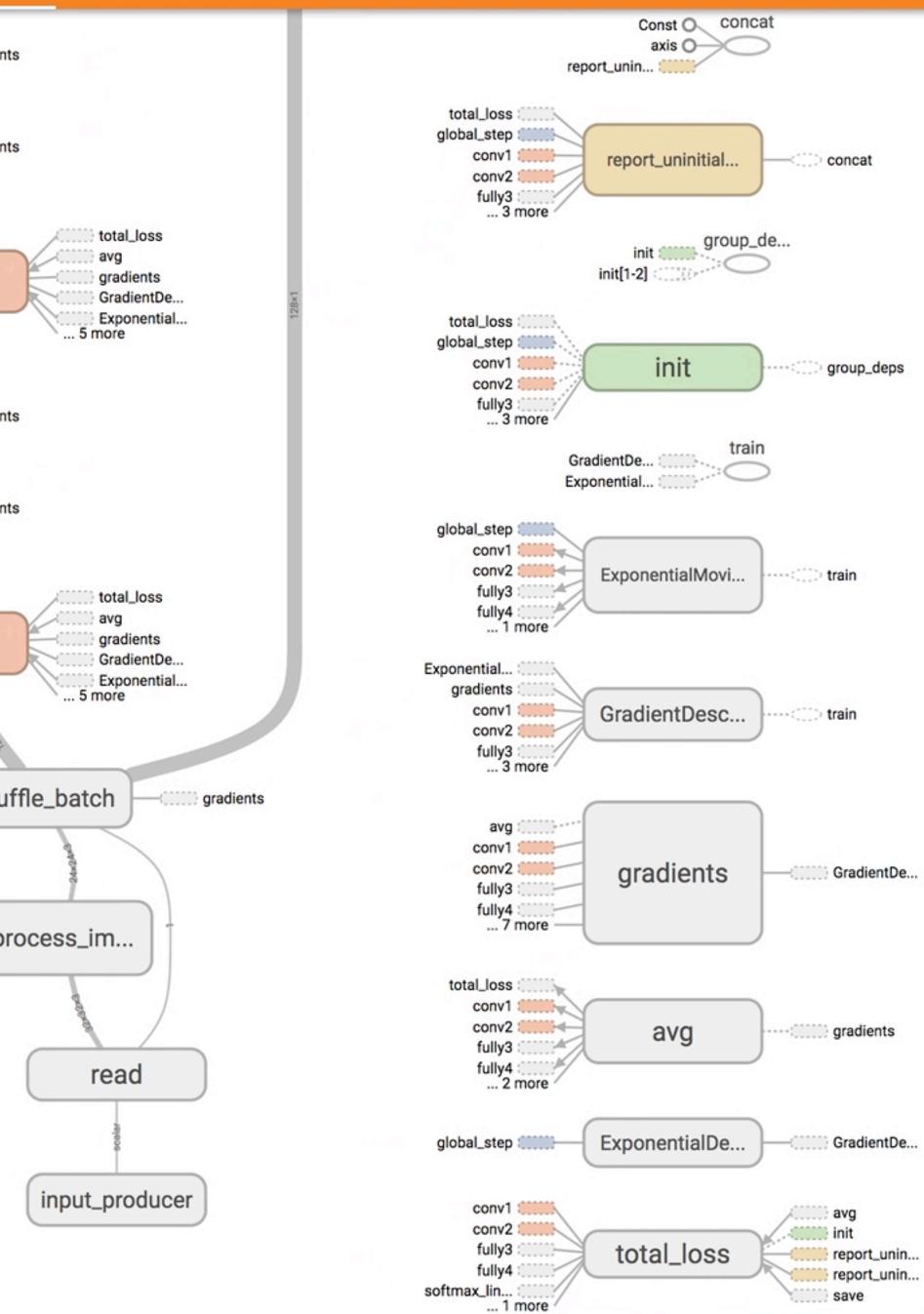


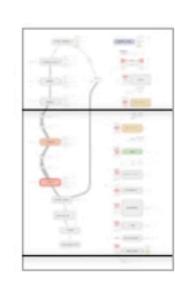
-

?

TensorBoard	SCALARS IMAGES A	UDIO	GRAPH
Fit to screen		pool2	
Download PNG		nofm2	
Run <u>run1 ~</u> (2)		Parauxon	gradient
Session	Identity[0-9]	_	sonv2
Upload Choose File			- Ange
Trace inputs		norm	gradient
Color 🔘 Structure		2×12×64	gradent
O Device		pool	
colors same substructure unique substructure		2446A	
		Abstate of	1000
	Identity[0-9]	-(conv1
			CANANG
			shu
	Modules		_
	Group of operations that	_	pr
Graph (* = expandable)	perform certain functions		
Namespace*			
OpNode			
Constant II. Summary			2
Dataflow edge			
Control dependency edg	e		
Reference edge			



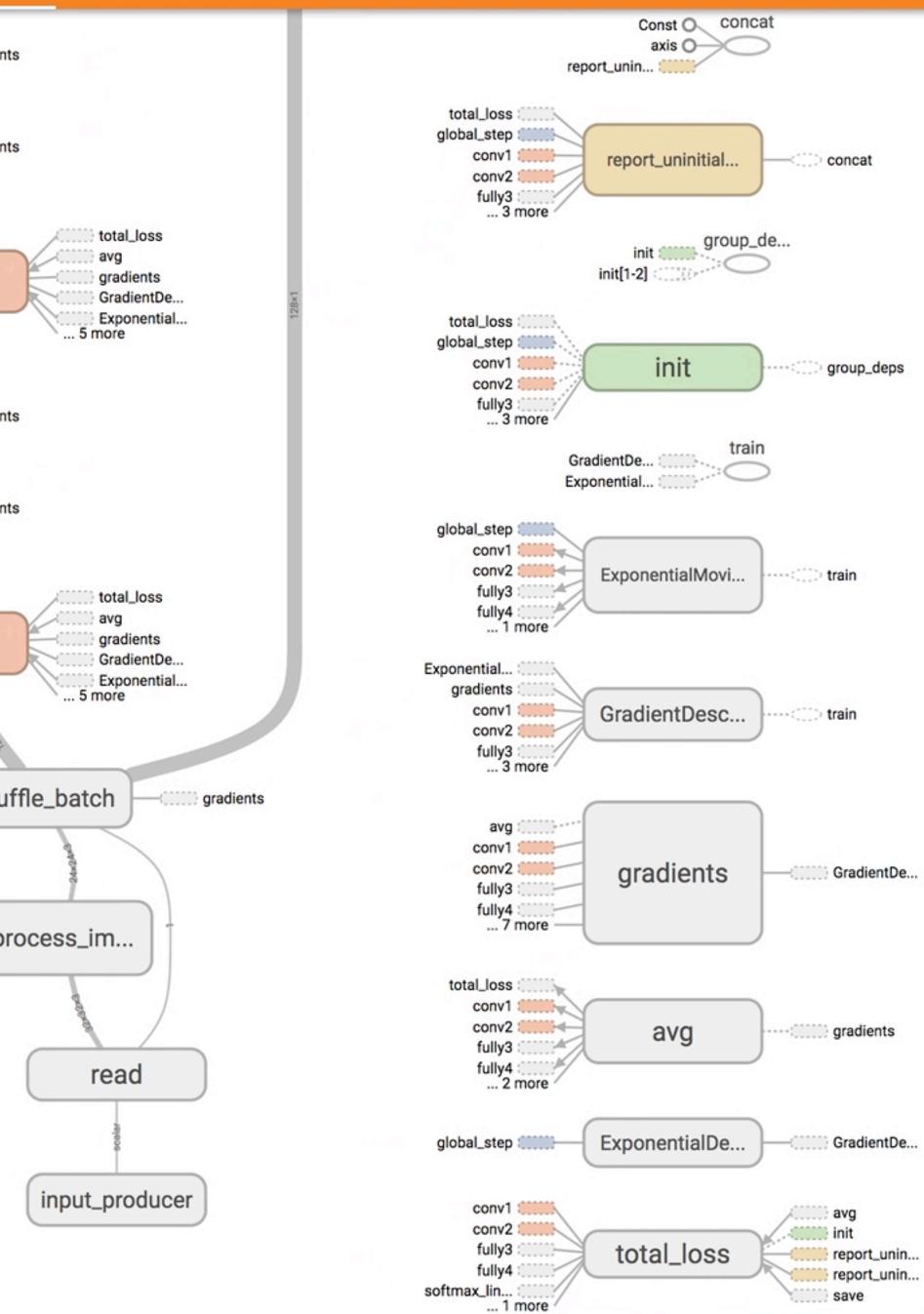


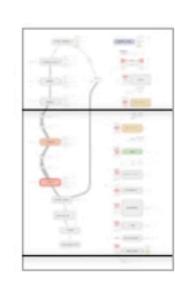




TensorBoard	SCALARS IMAGES A	UDIO	GRAPH
Fit to screen		pool2	
Download PNG		nofm2	
Run <u>run1 ~</u> (2)		Parauxon	gradient
Session	Identity[0-9]	_	sonv2
Upload Choose File			- Ange
Trace inputs		norm	gradient
Color 🔘 Structure		2×12×64	gradent
O Device		pool	
colors same substructure unique substructure		2446A	
		Abstate of	1000
	Identity[0-9]	-(conv1
			CANAND
			shu
	Modules		_
	Group of operations that	_	pr
Graph (* = expandable)	perform certain functions		
Namespace*			
OpNode			
Constant II. Summary			2
Dataflow edge			
Control dependency edg	e		
Reference edge			









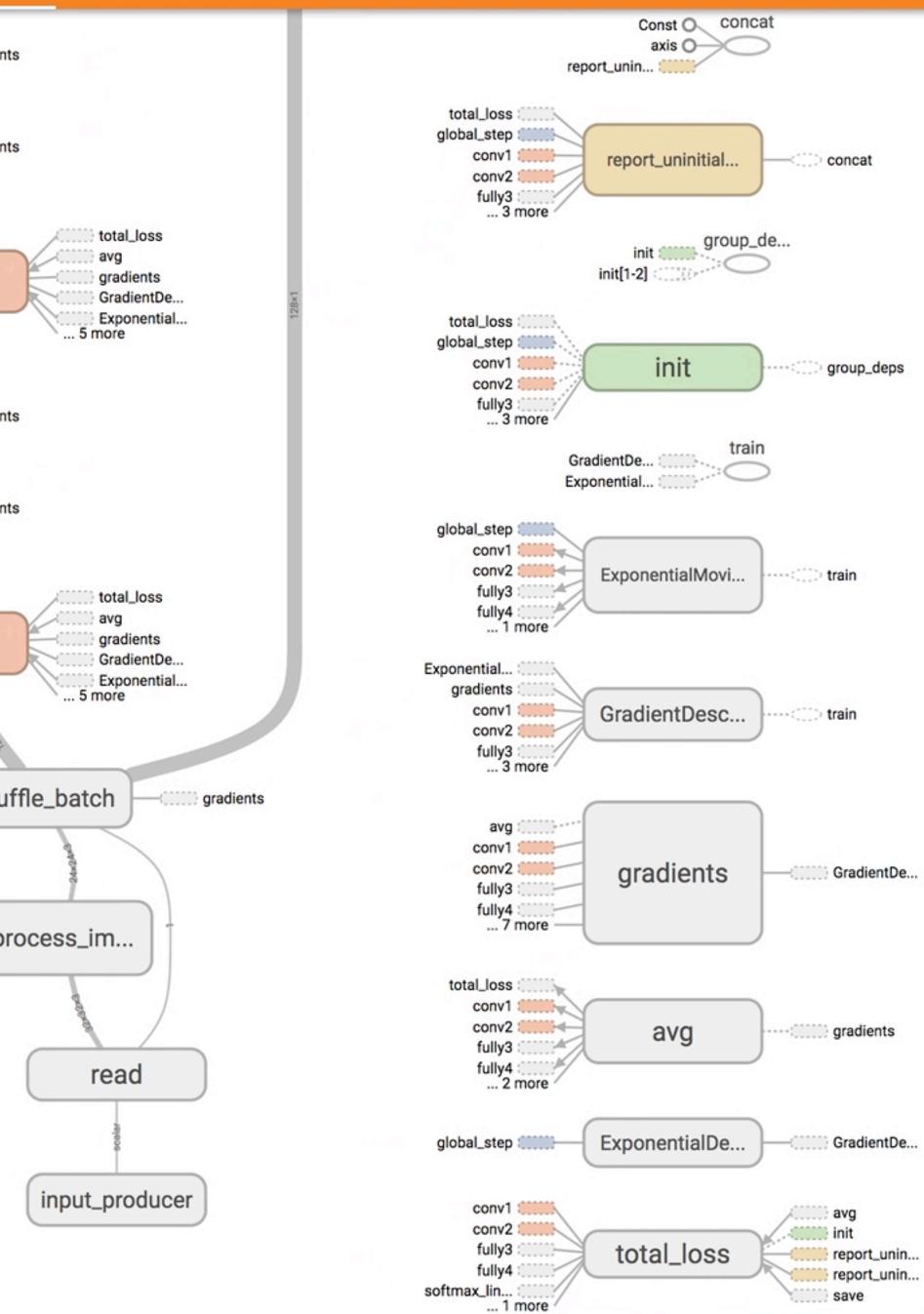
TensorBoard	SCALARS	IMAGES	AUDIO	GRAPHS
Fit to screen			2x12x64	gradients
Download PNG Run <u>run1 ~</u> (2)			notin2	gradients
Session runs (0)		Identity[0-9]	ivau:	conv2
Upload Choose File			P12000	
Trace inputs	Individual		norm	
Color () Structure	muividuai		2×12×64	⊰) gradients
O Device colors same substructure	Operations		pool1	gradients
unique substructure			ABADARD MARCH	4
		Identity[0-9]	e-((conv1
			1	CANTON CAST
		•		shuf
	Мо	dule	S	
G	roup of operatio	ons tha	nt —	-> pro
pe	erform certain fu	nction	S	
Graph (* = expandable)				
OpNode				
O Constant			k	•
II. Summary				

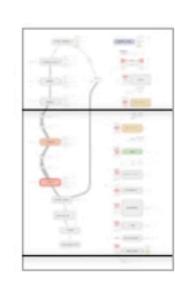
 Control	dependency	edge

Dataflow edge

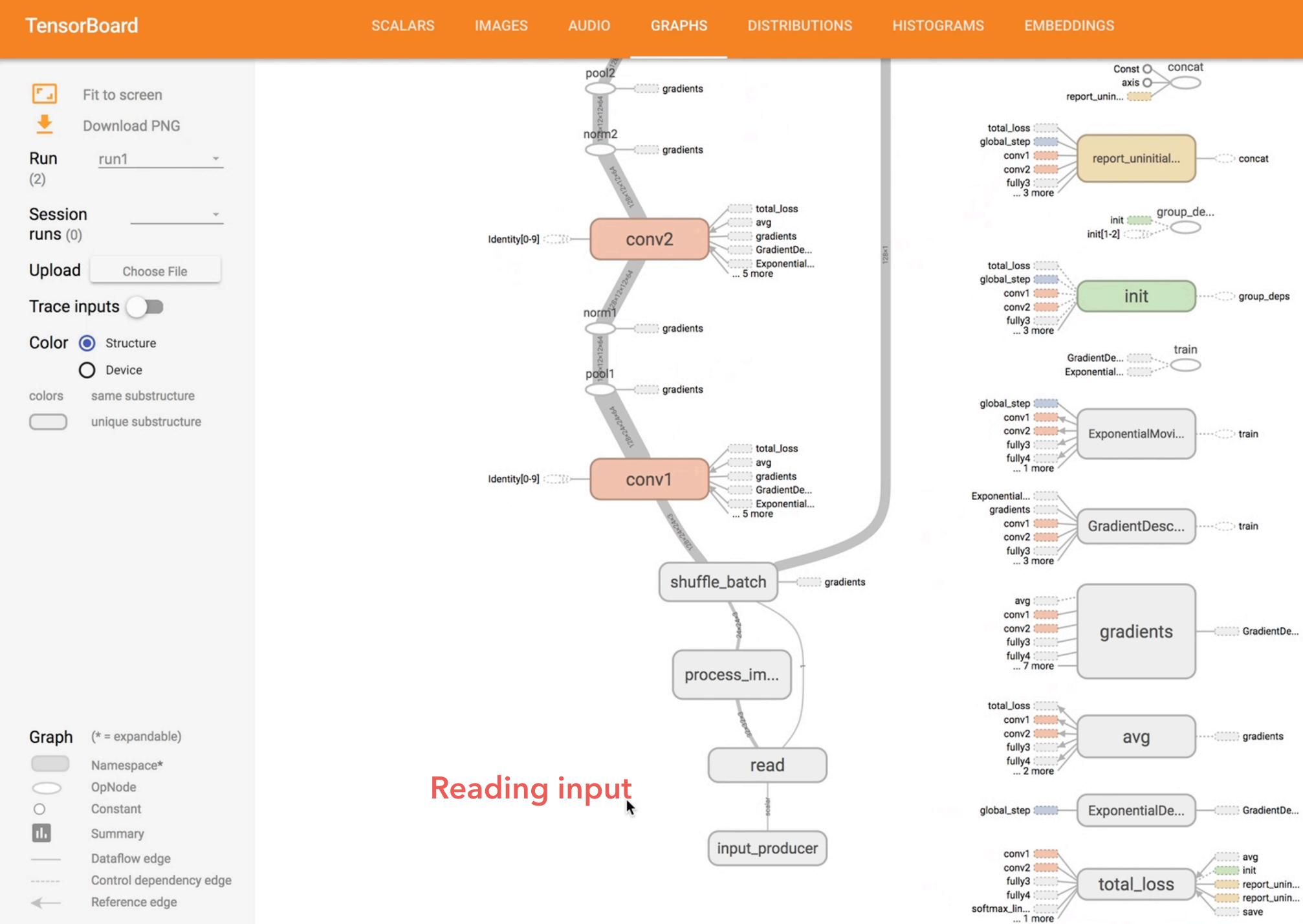
Reference edge



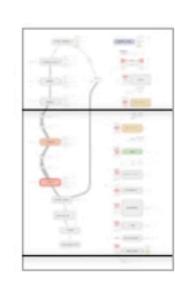




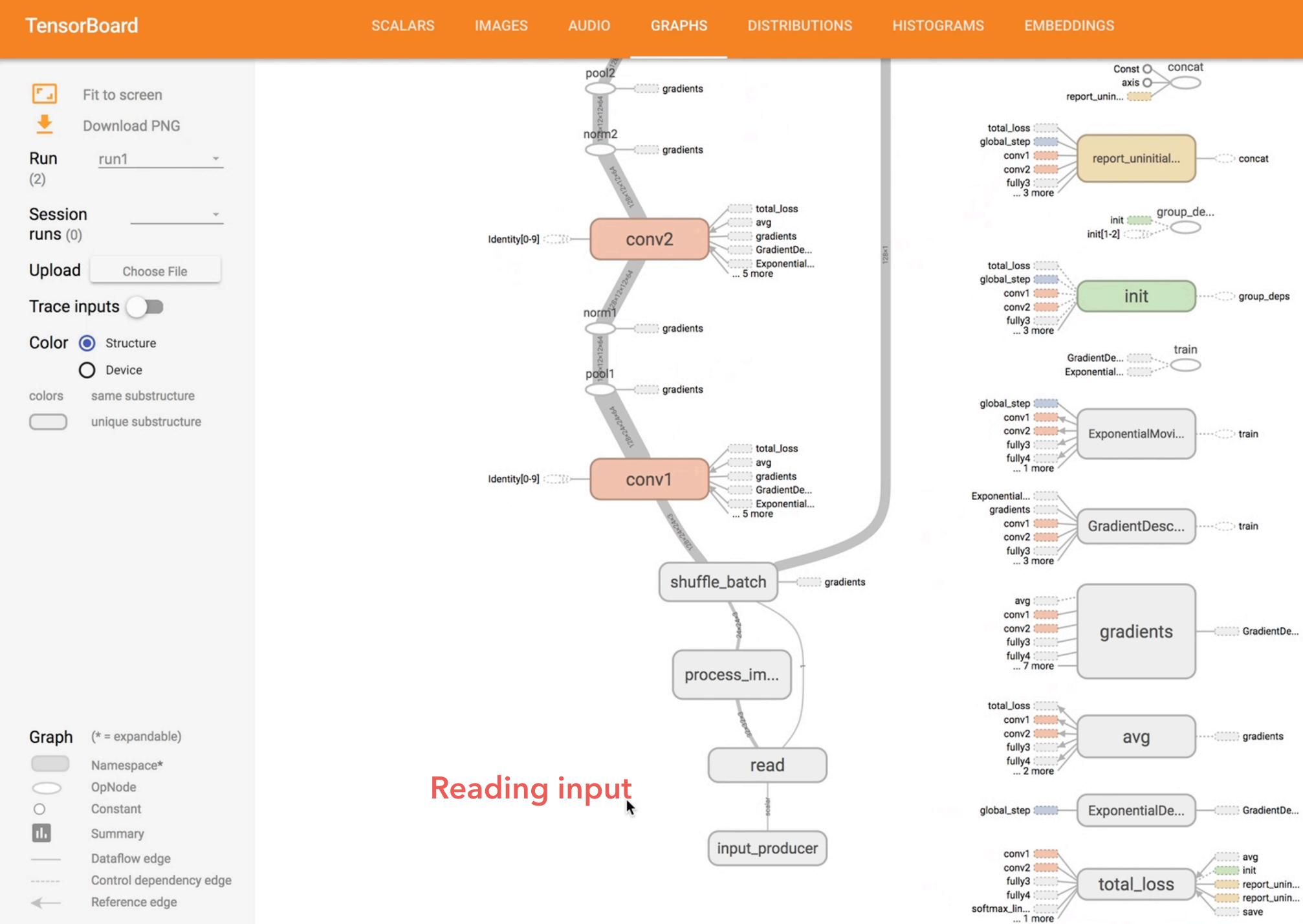




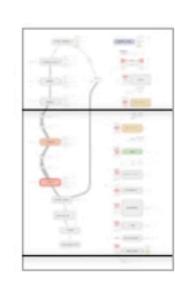








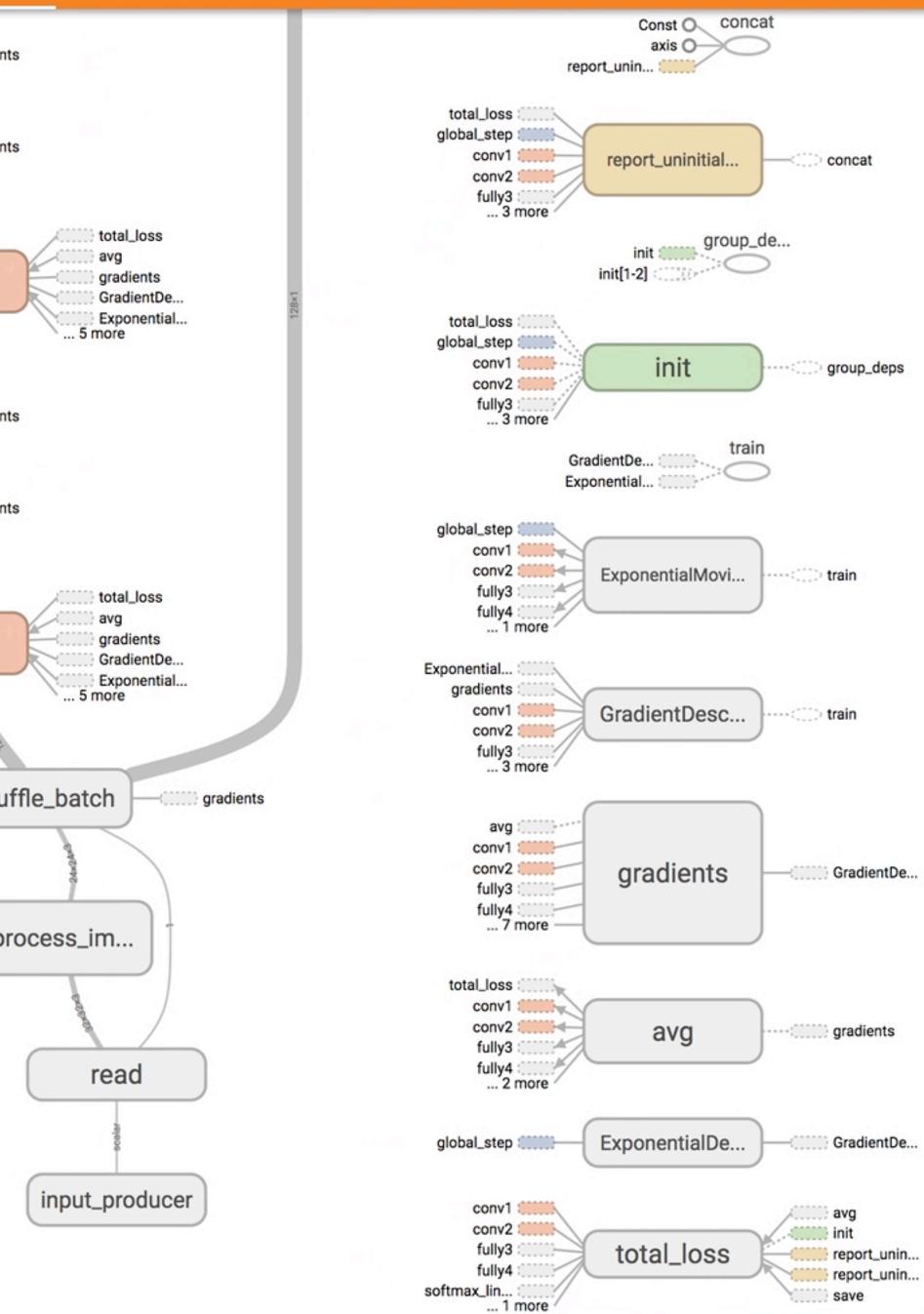


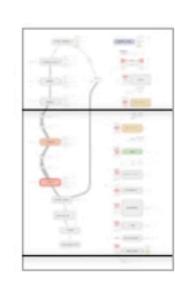




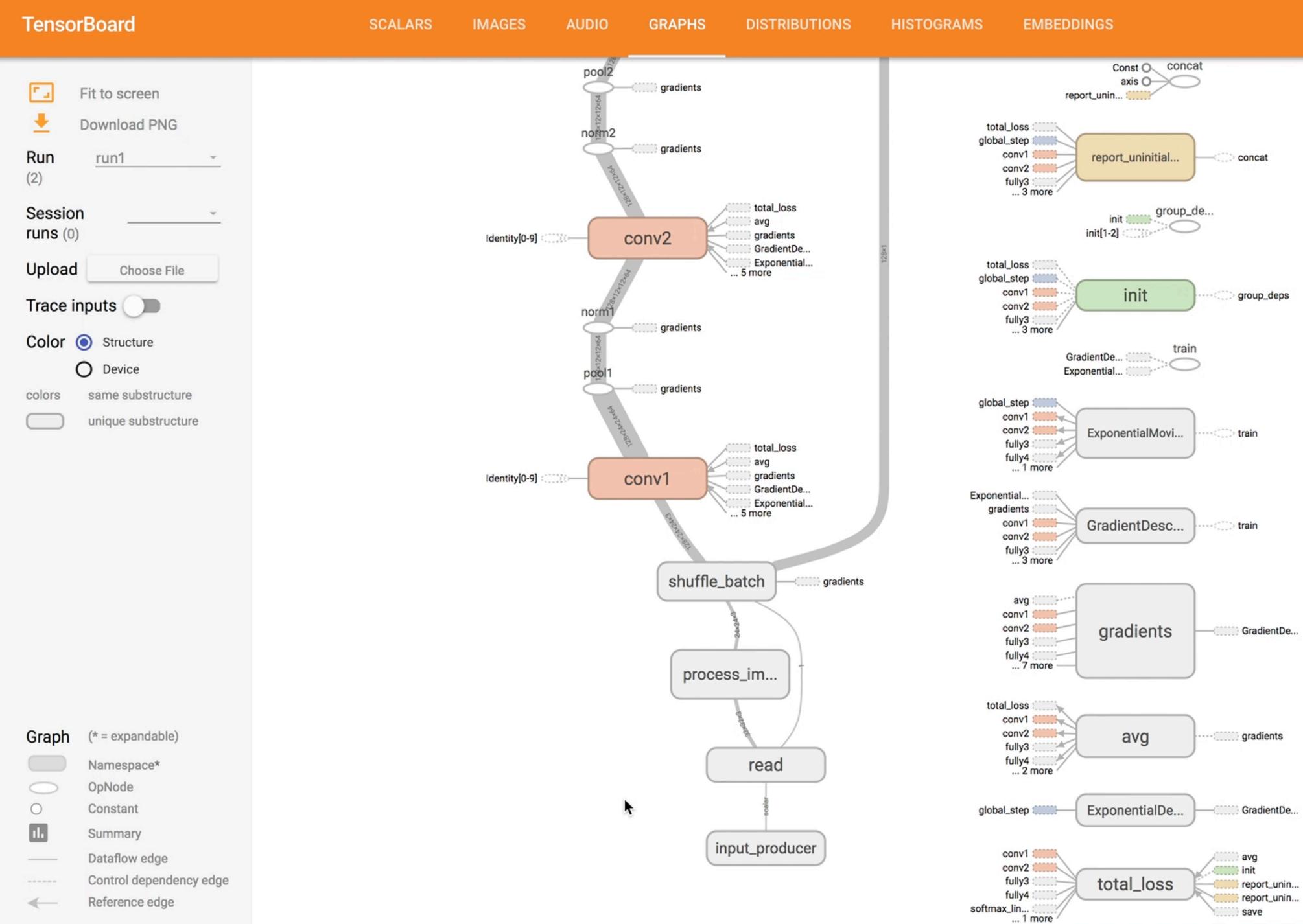
TensorBoard	SCALARS IMAGES AUDIO GRAPH
 Fit to screen Download PNG Run run1 · (a) Session · Choose File Upload Choose File Upload Choose File One on Structure O Device colors same substructure inique substructure 	Identity[0-9] Conv2
	Apply image processing (produce additional data)
Graph (* = expandable) Namespace* OpNode Constant	Reading input
Summary Dataflow edge Control dependency edge Reference edge	



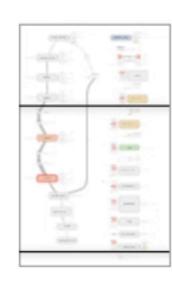




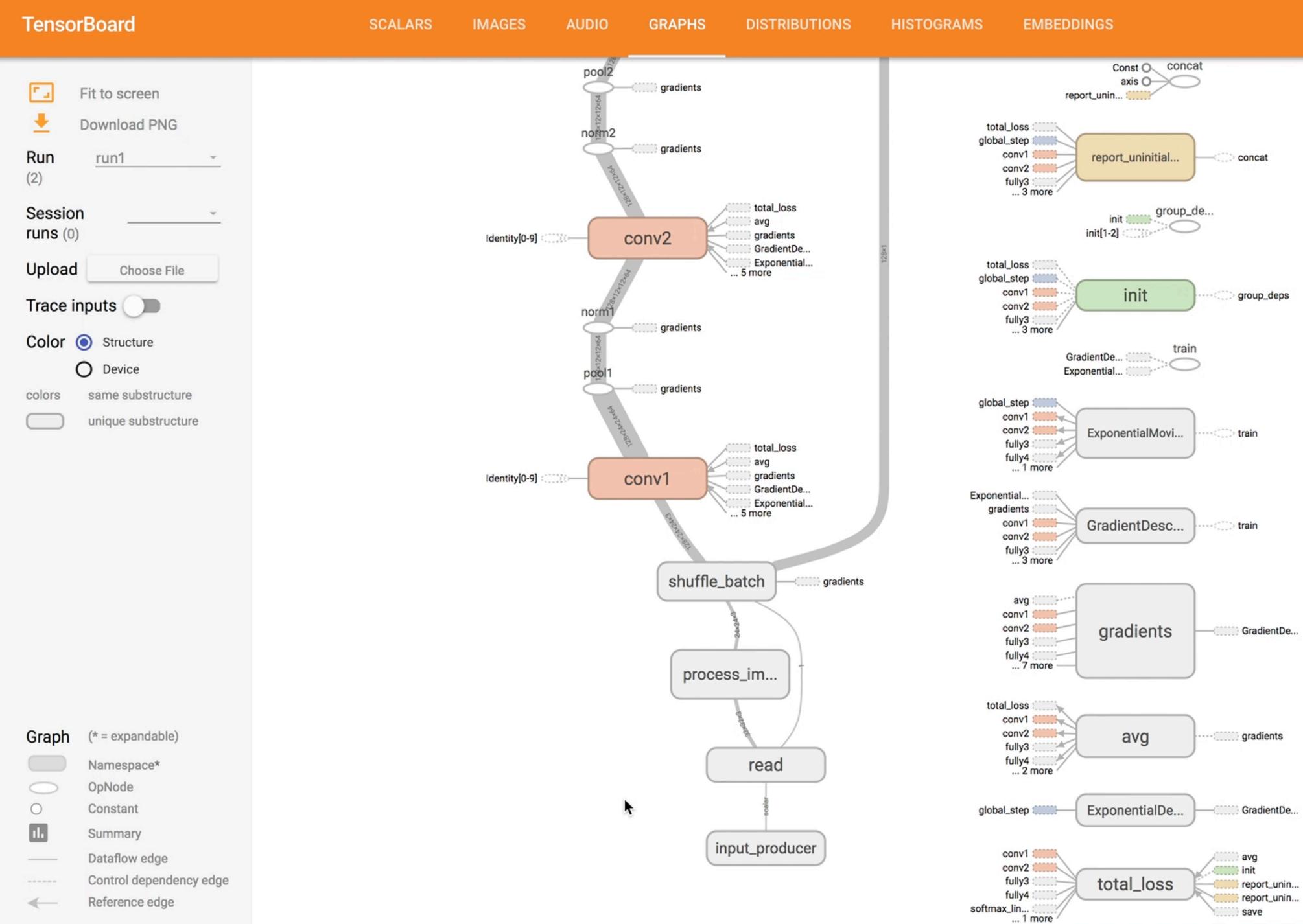




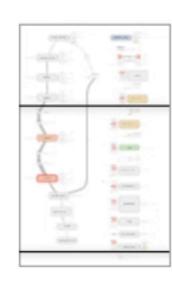




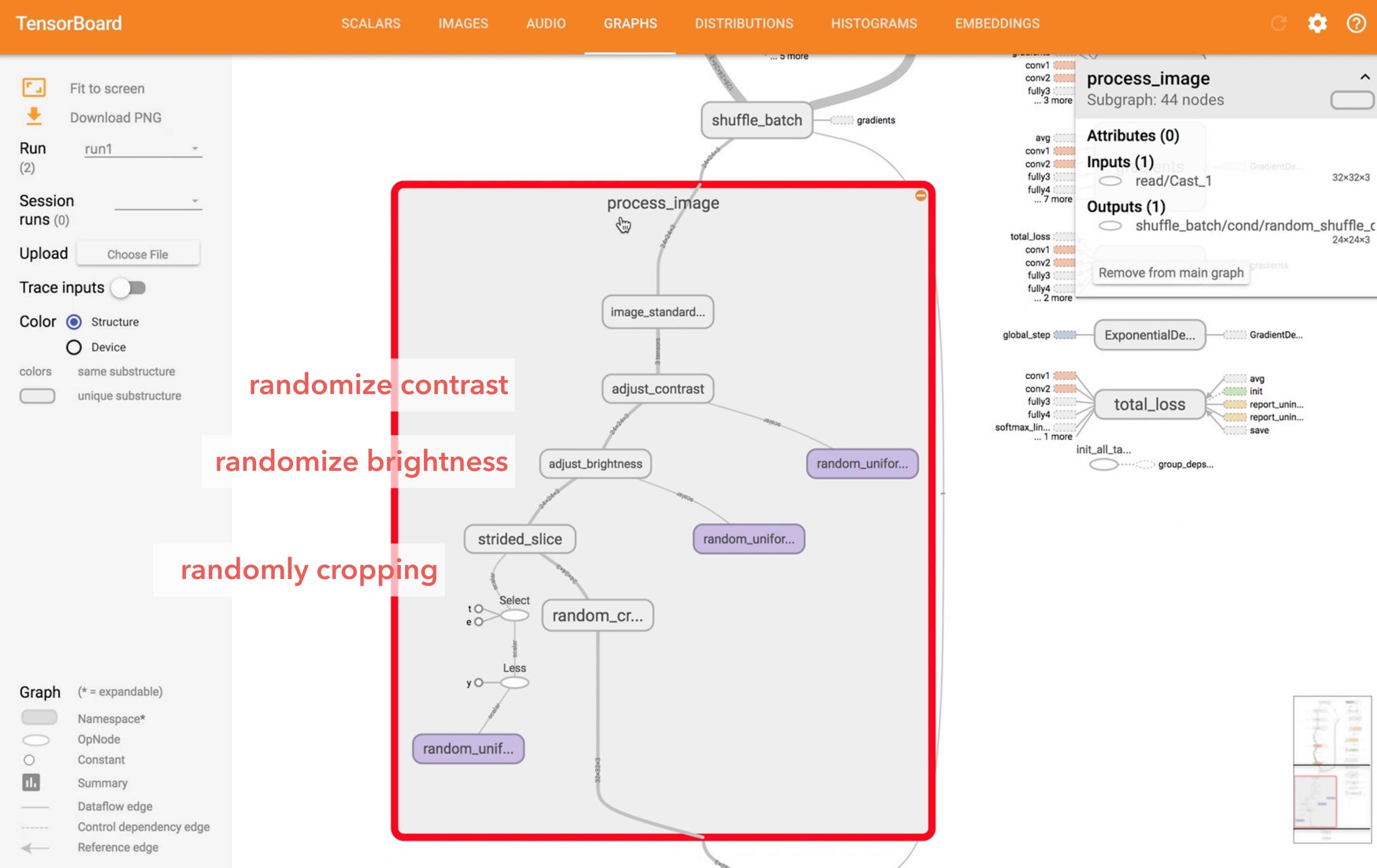








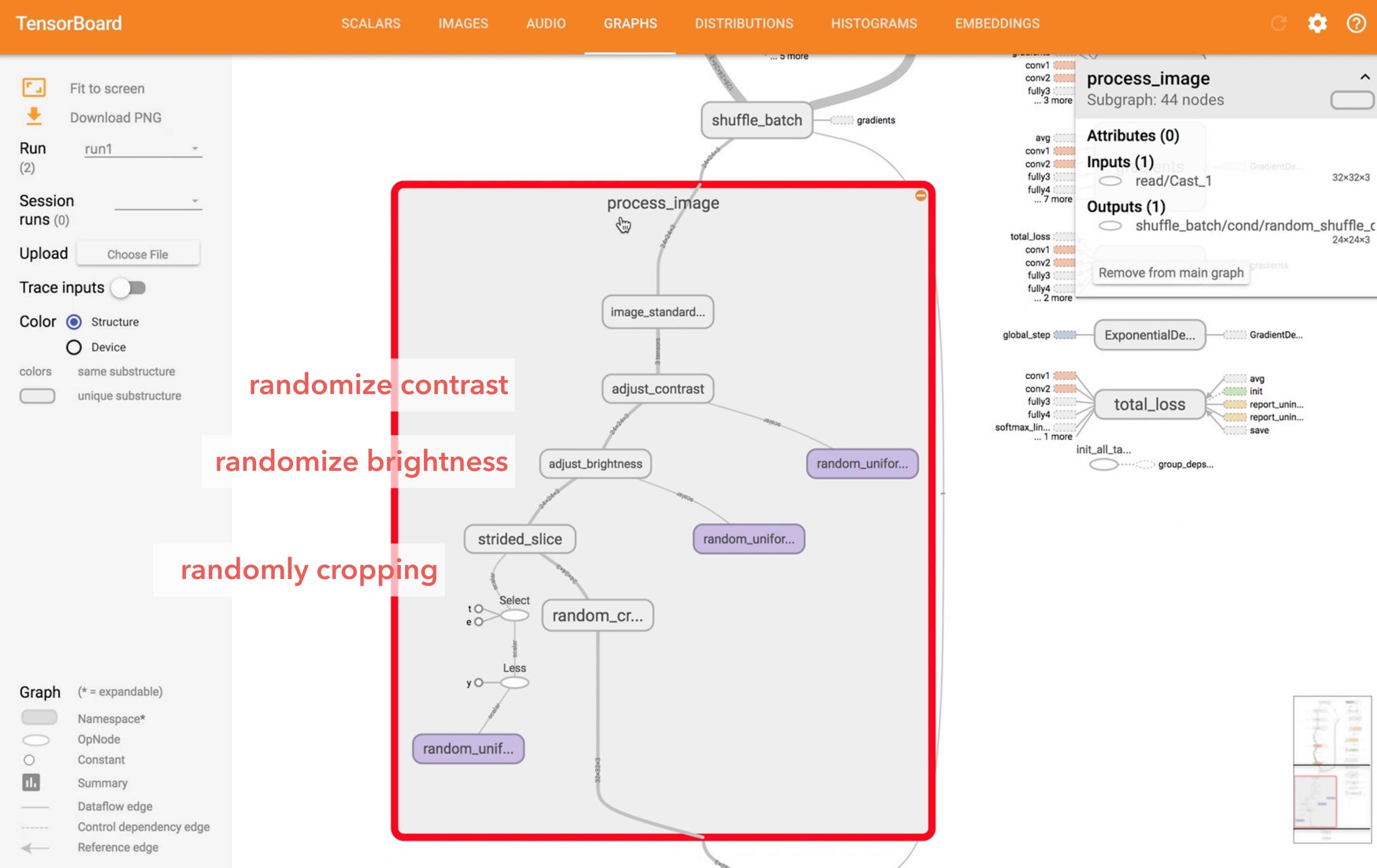








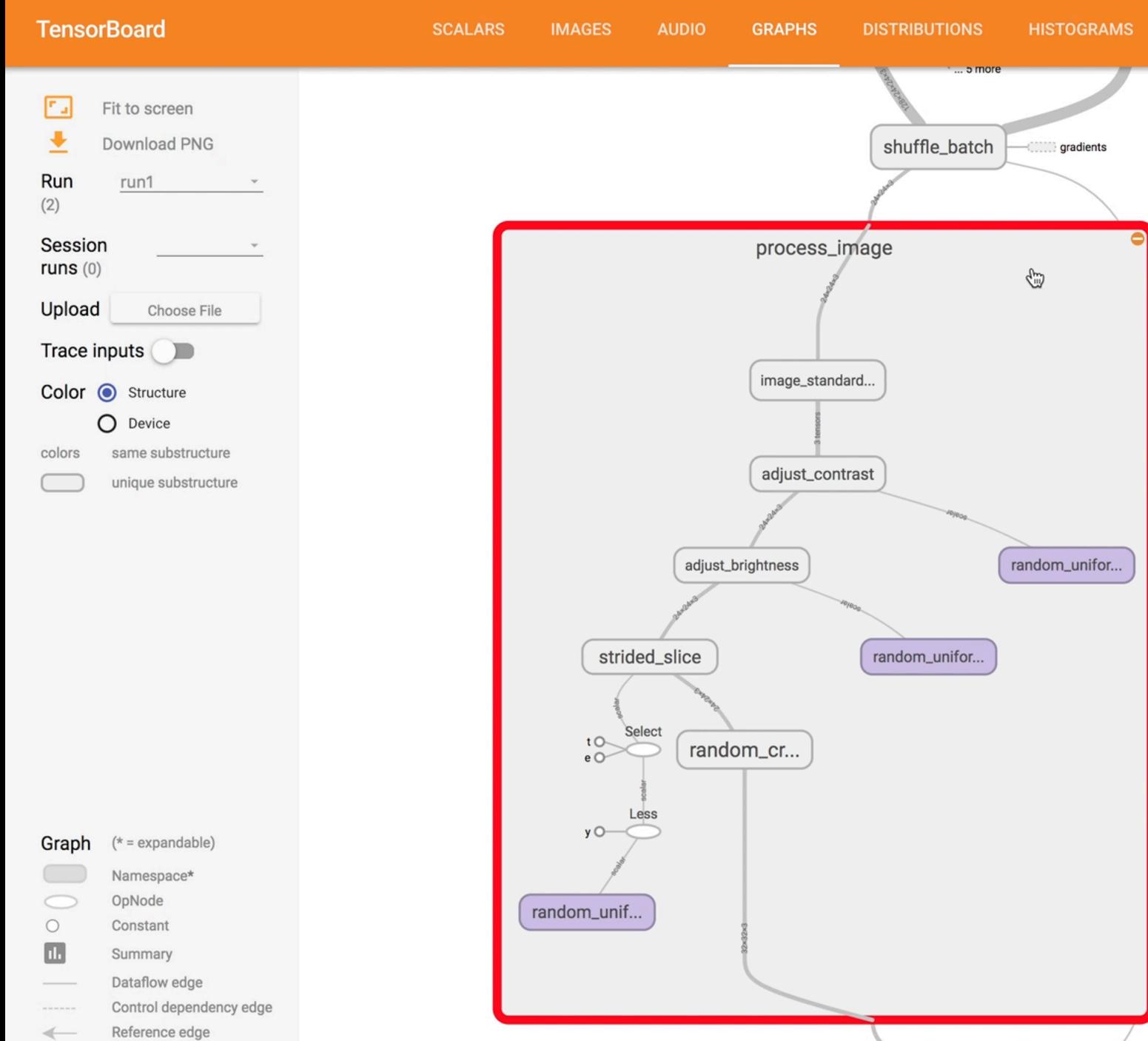




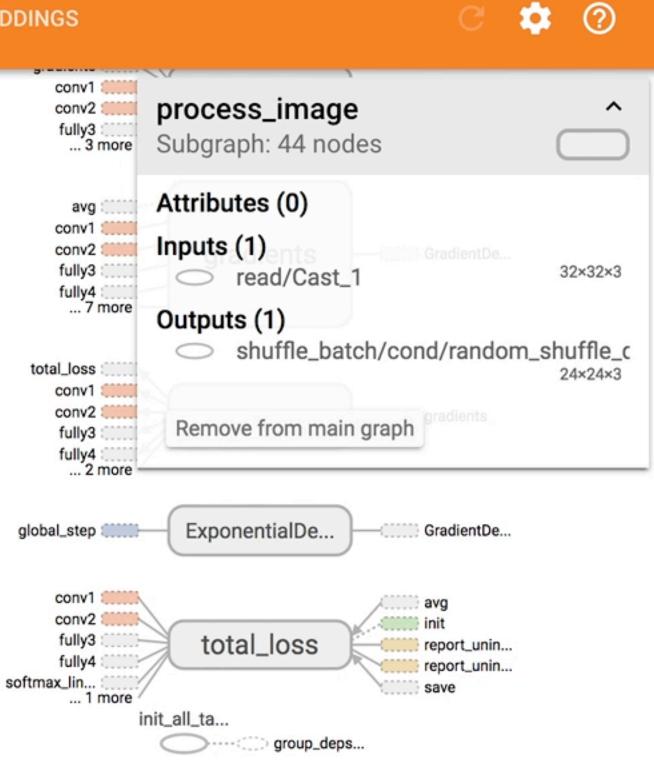


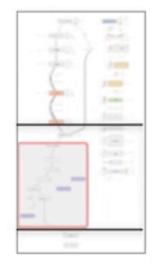


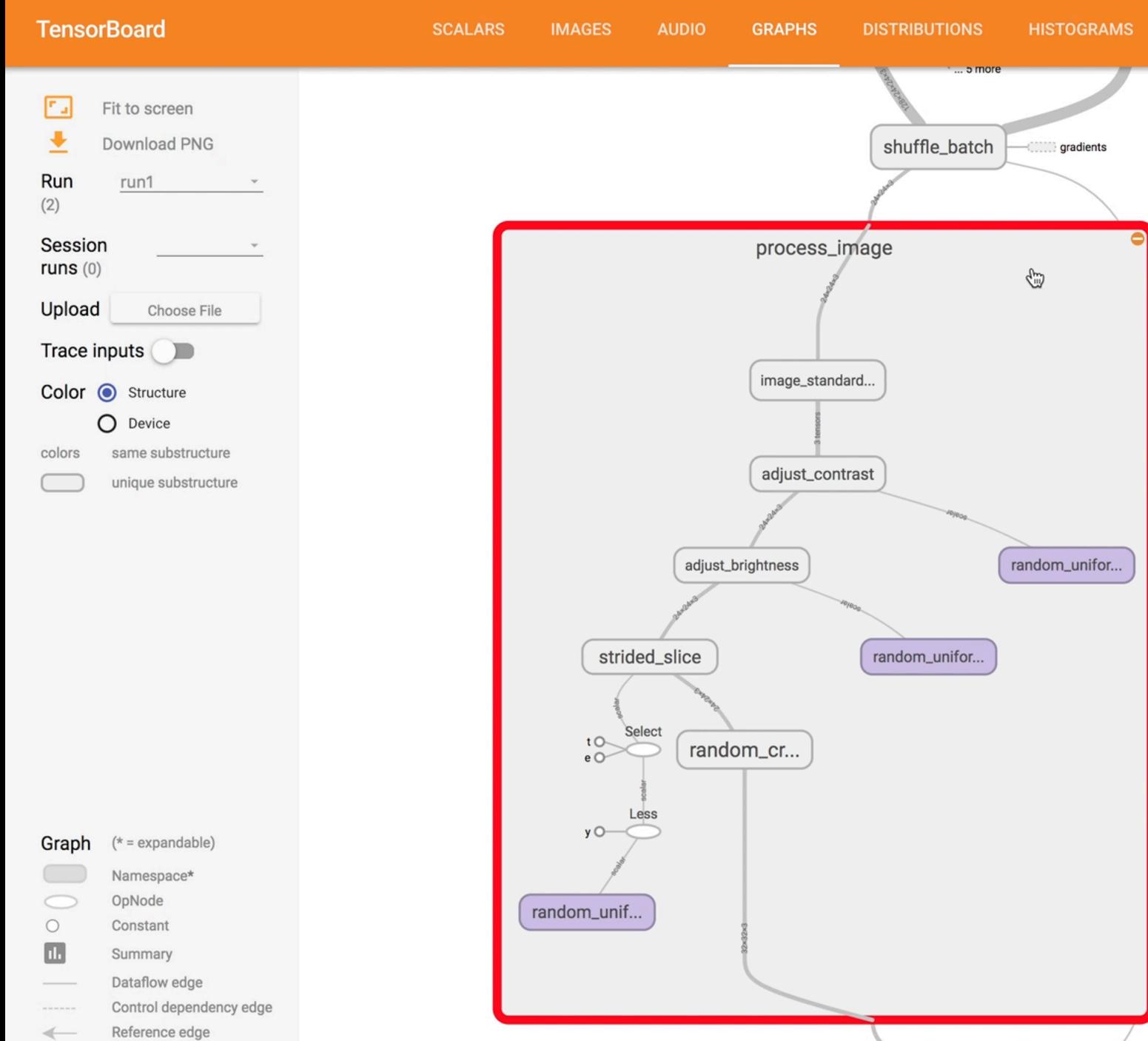




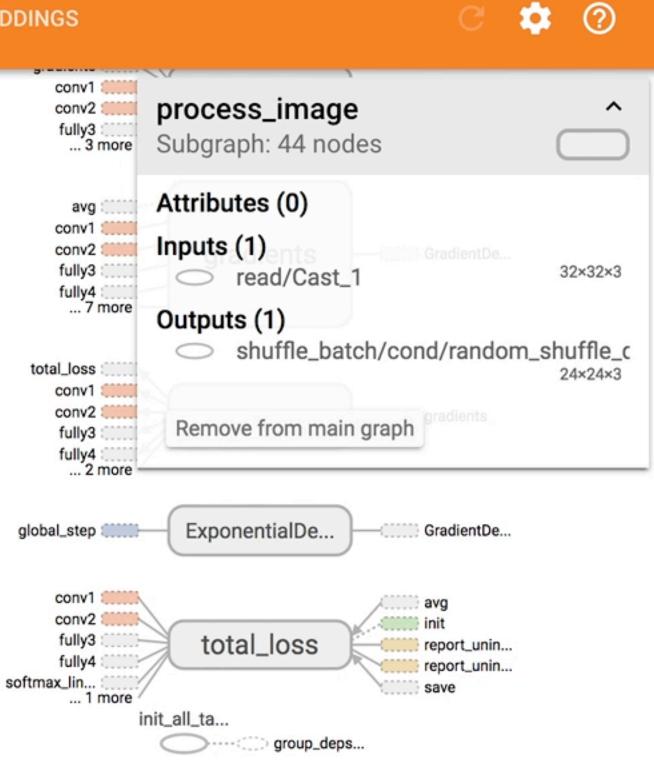
EMBEDDINGS

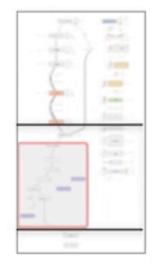


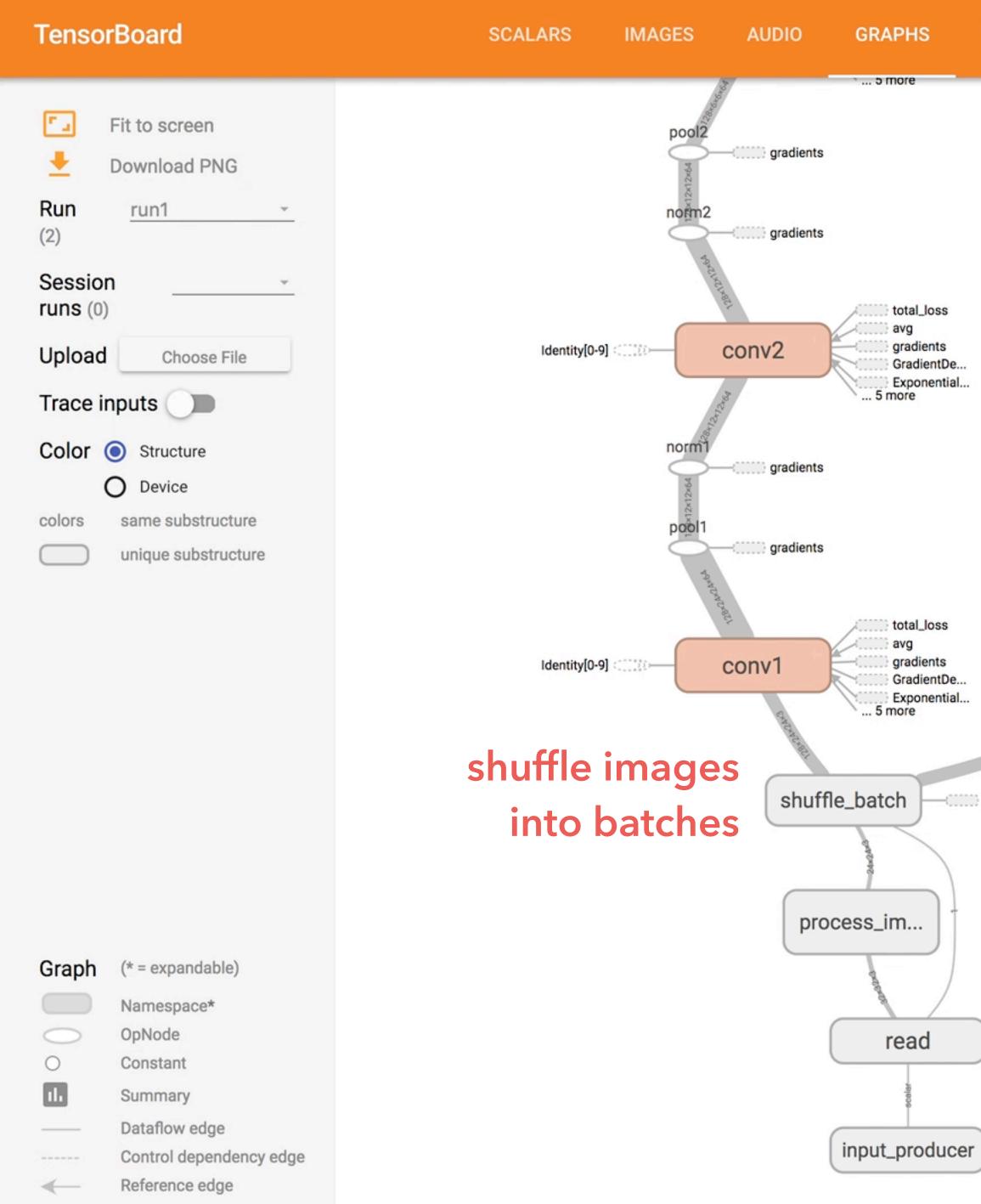




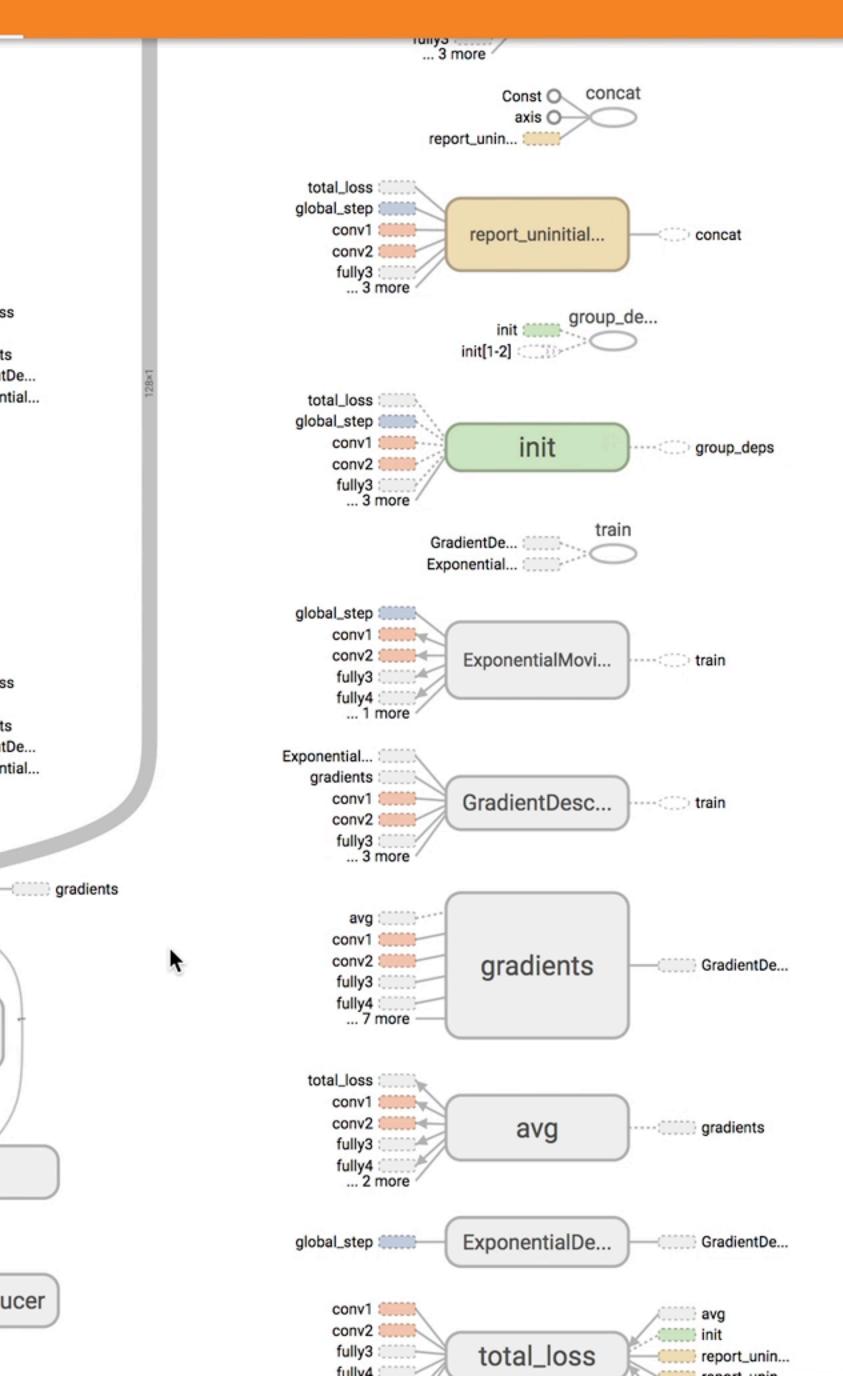
EMBEDDINGS

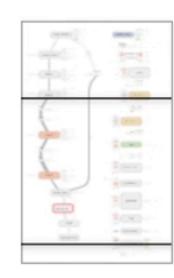




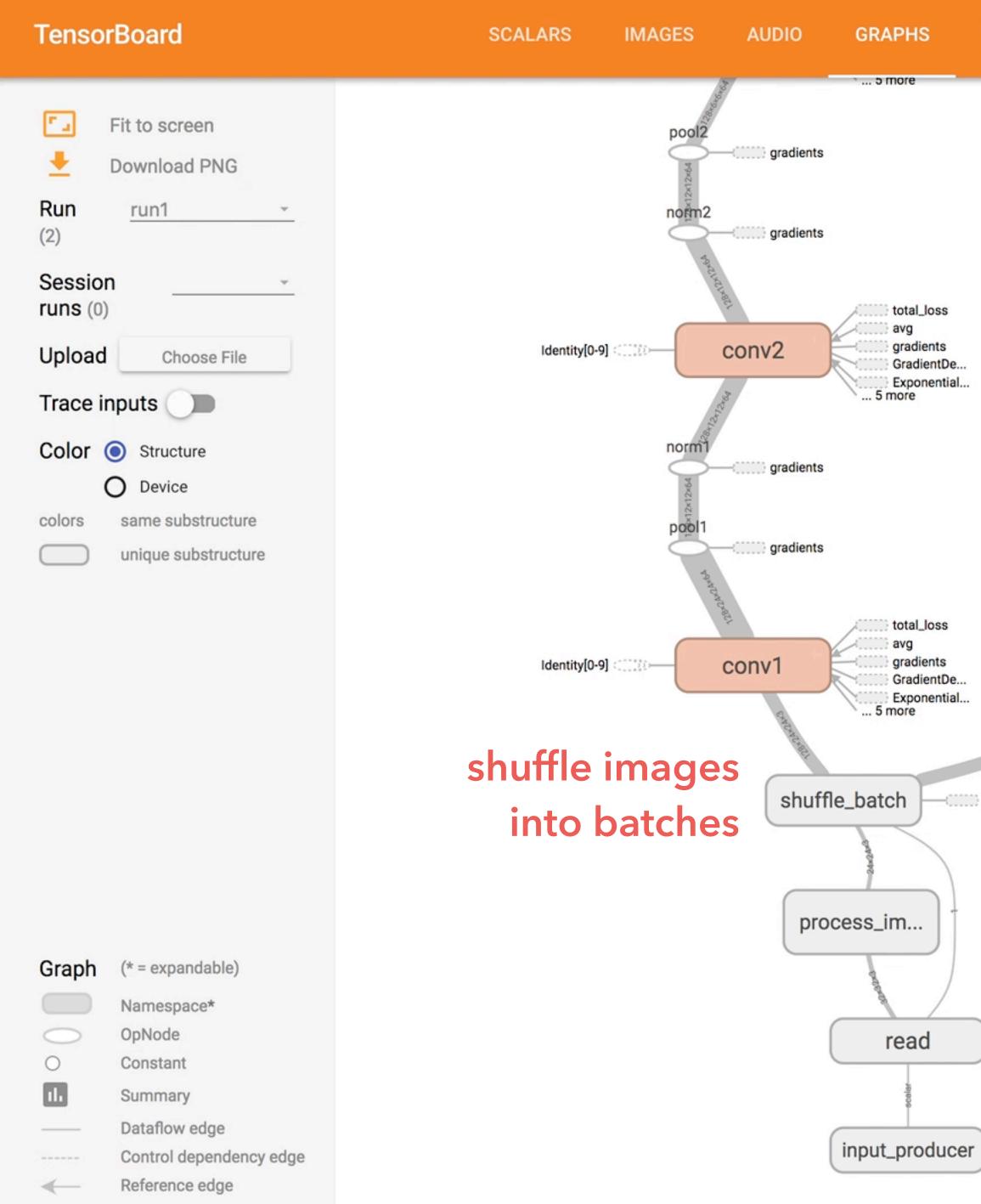




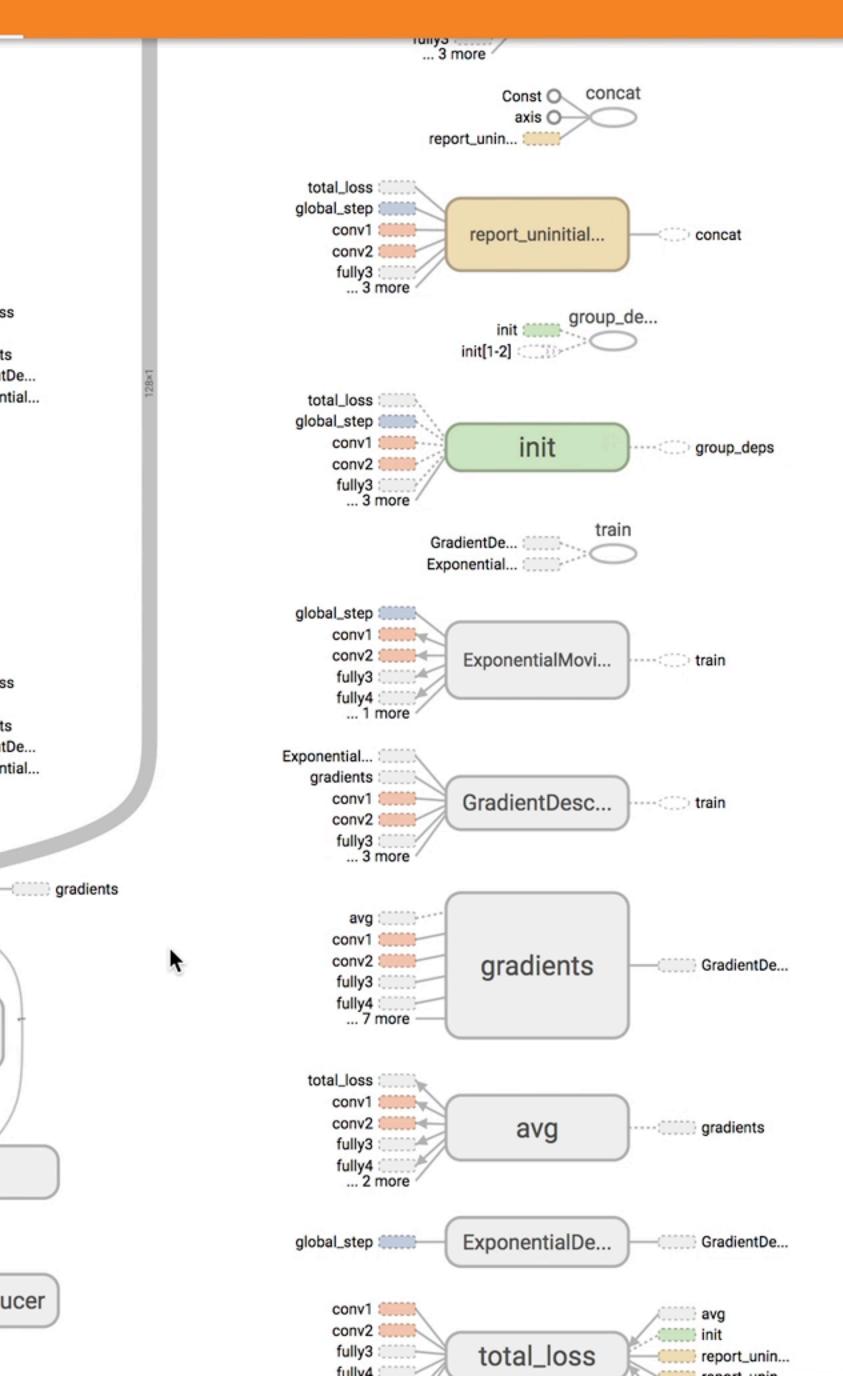


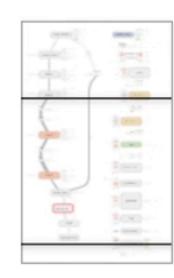




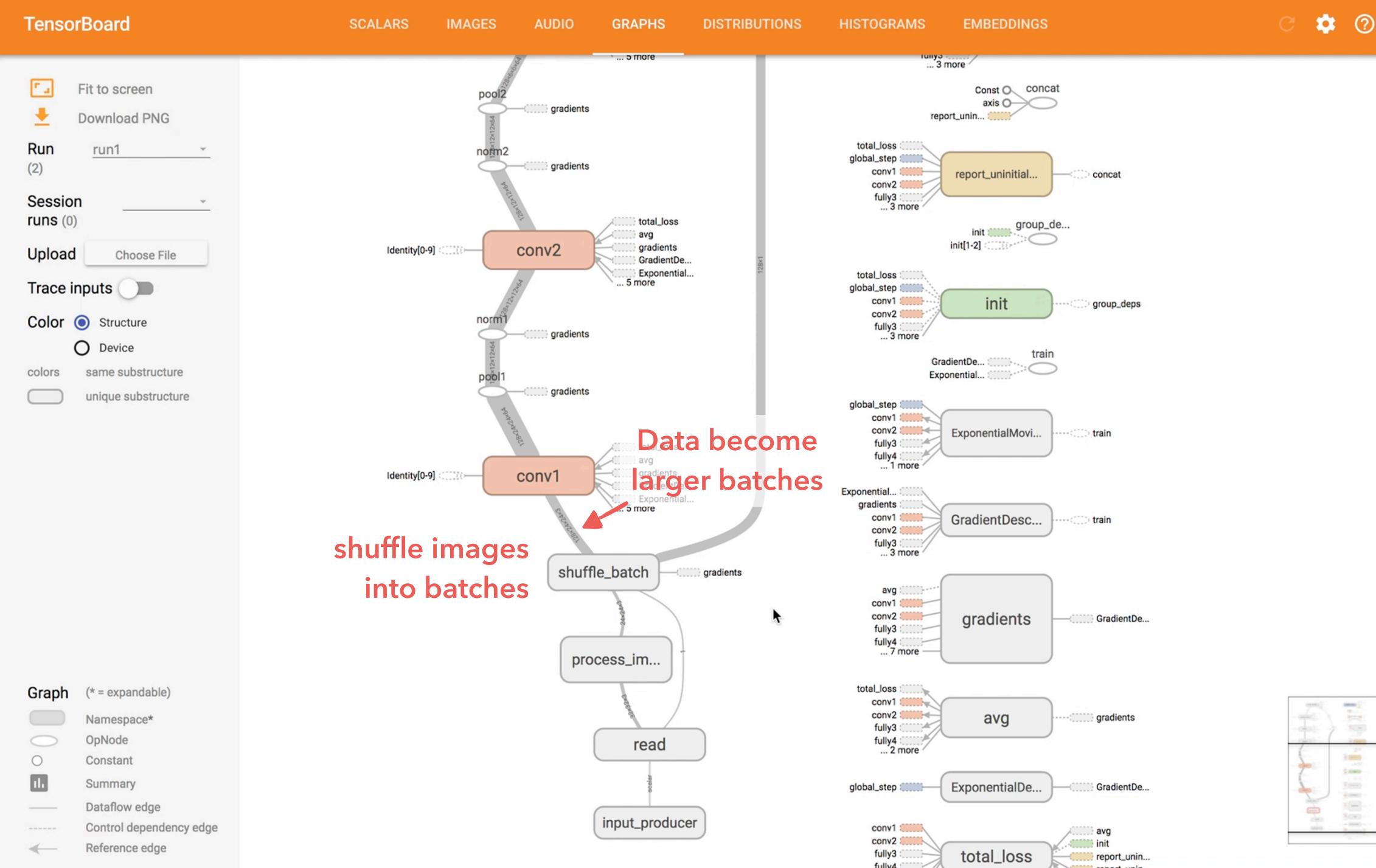




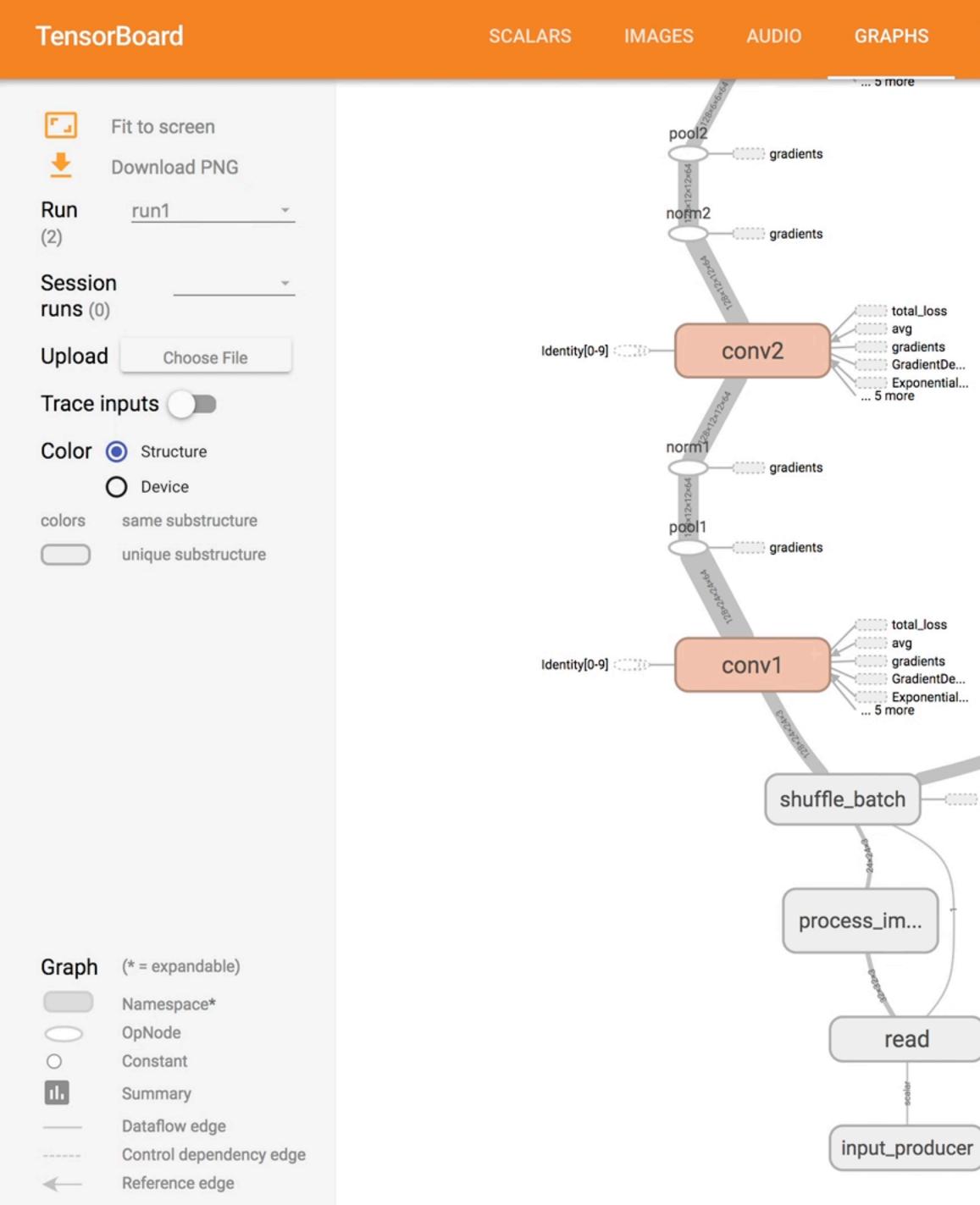




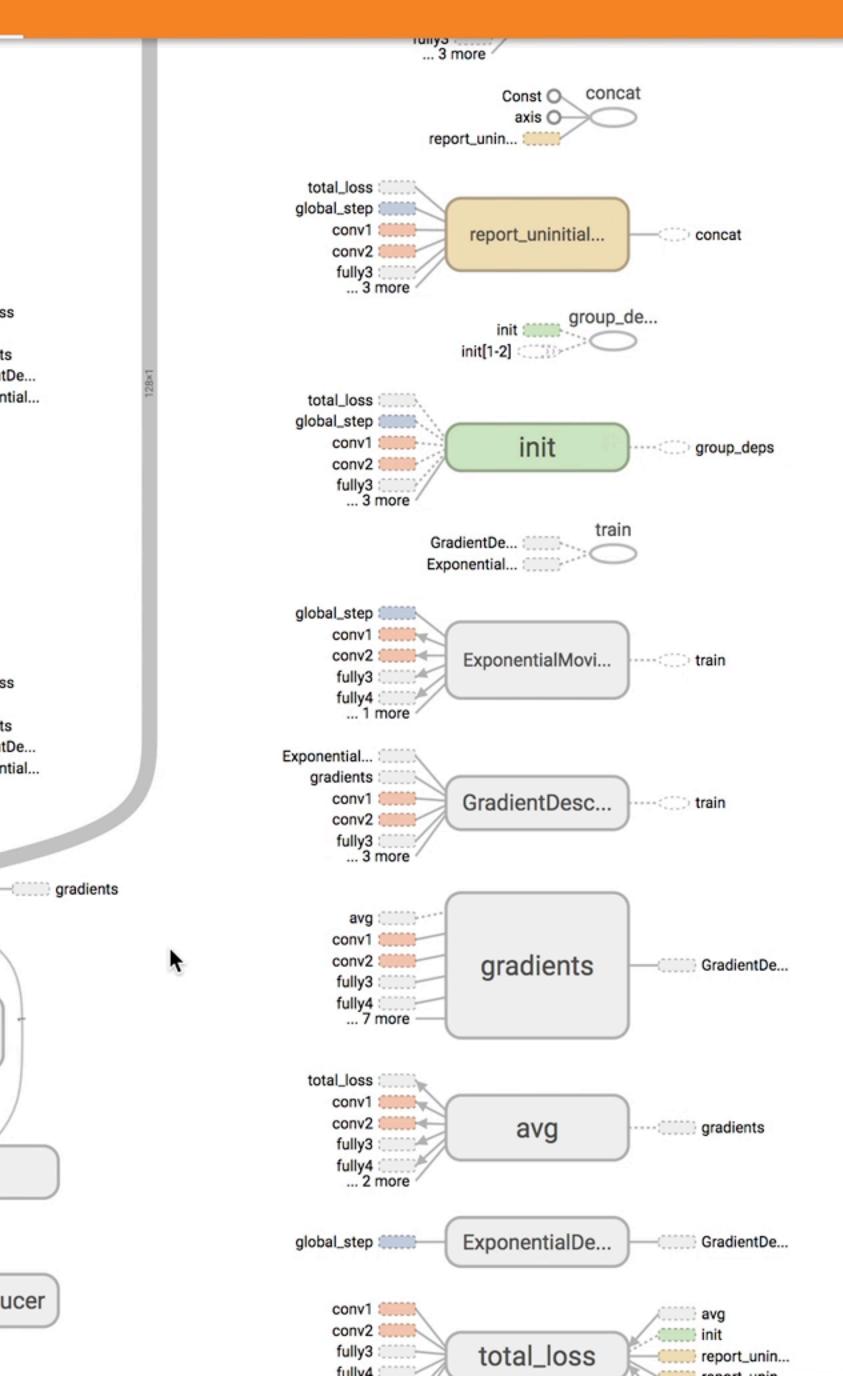


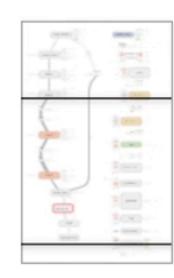




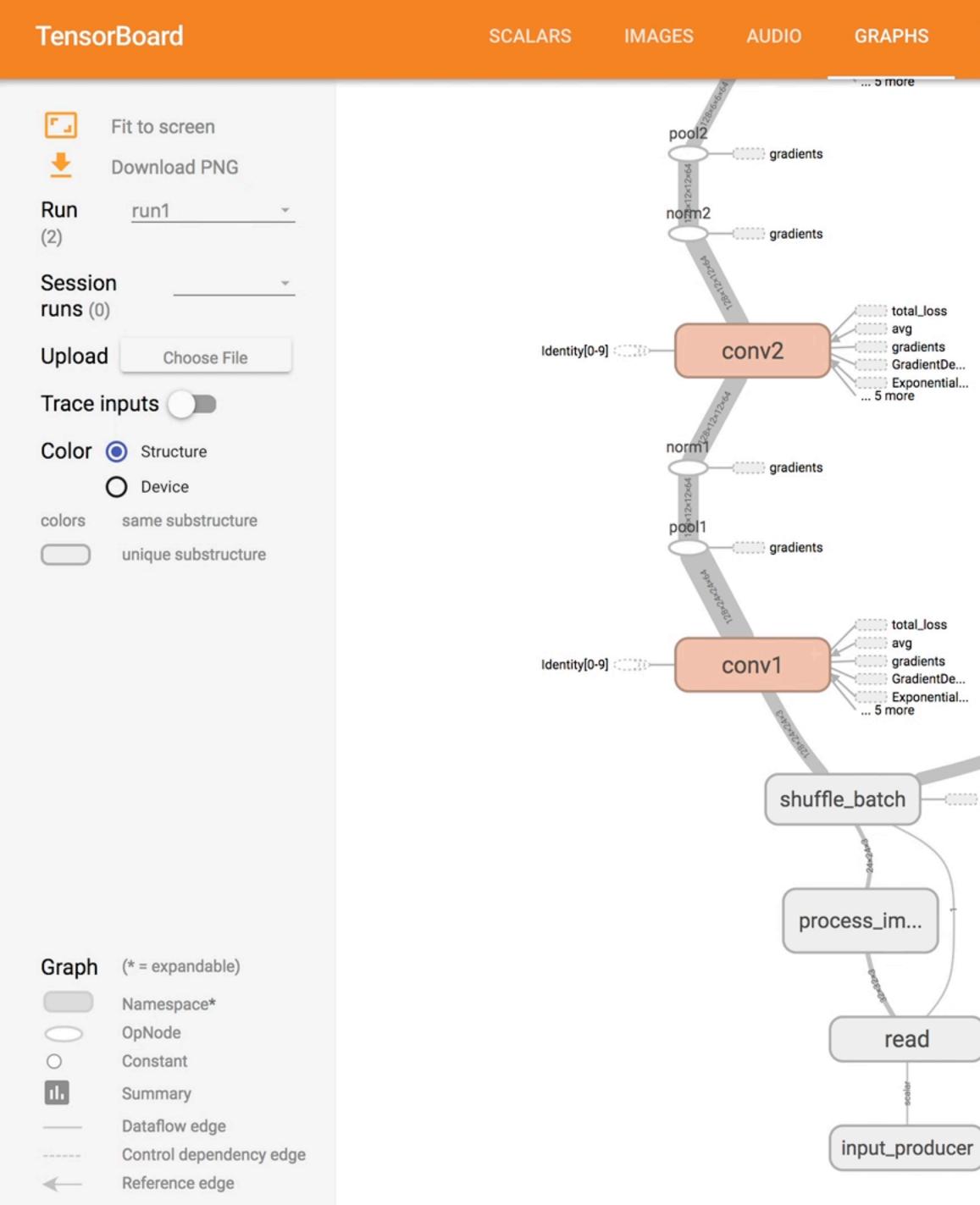




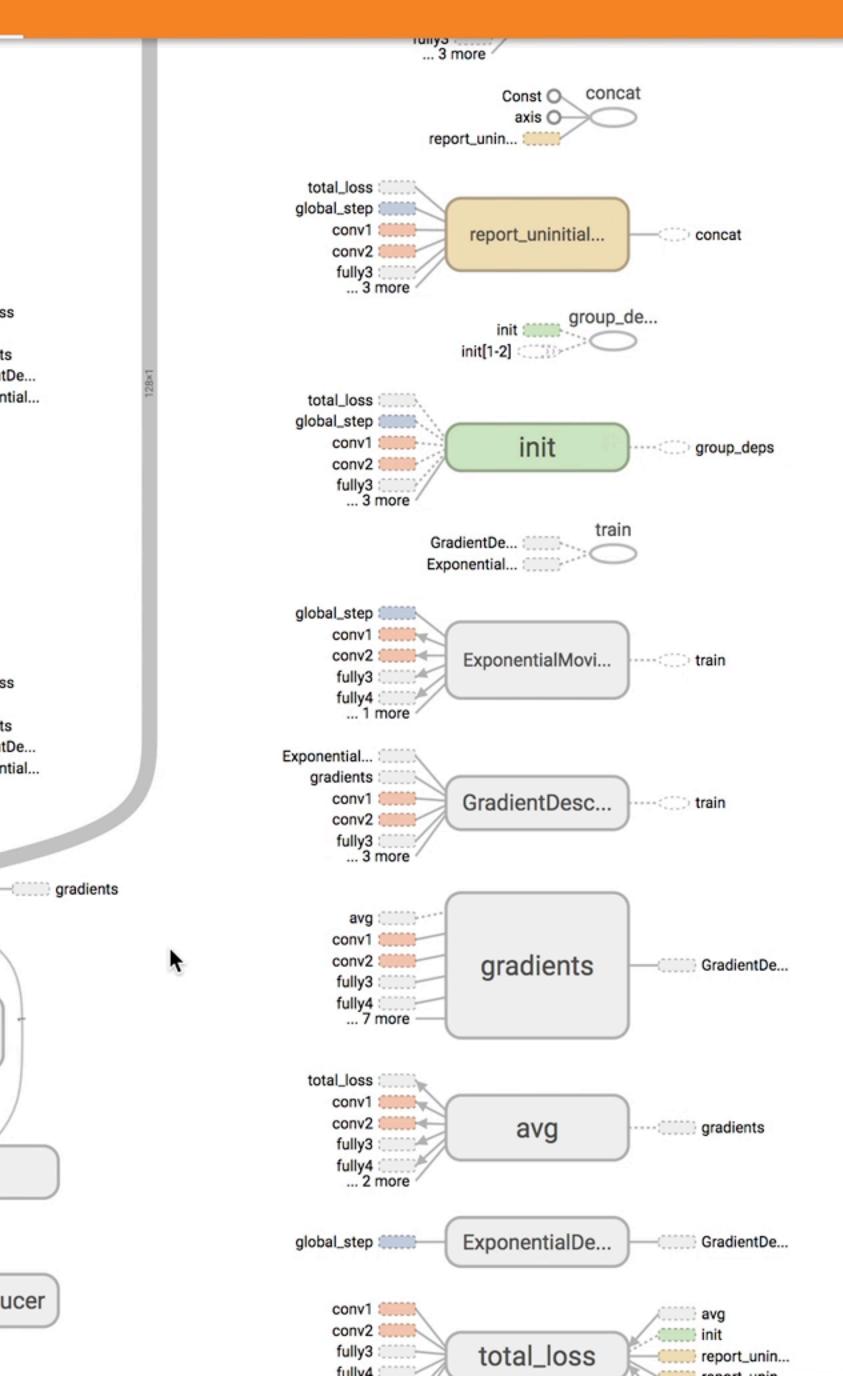


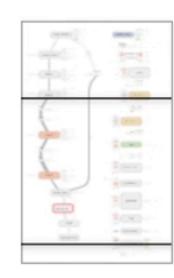




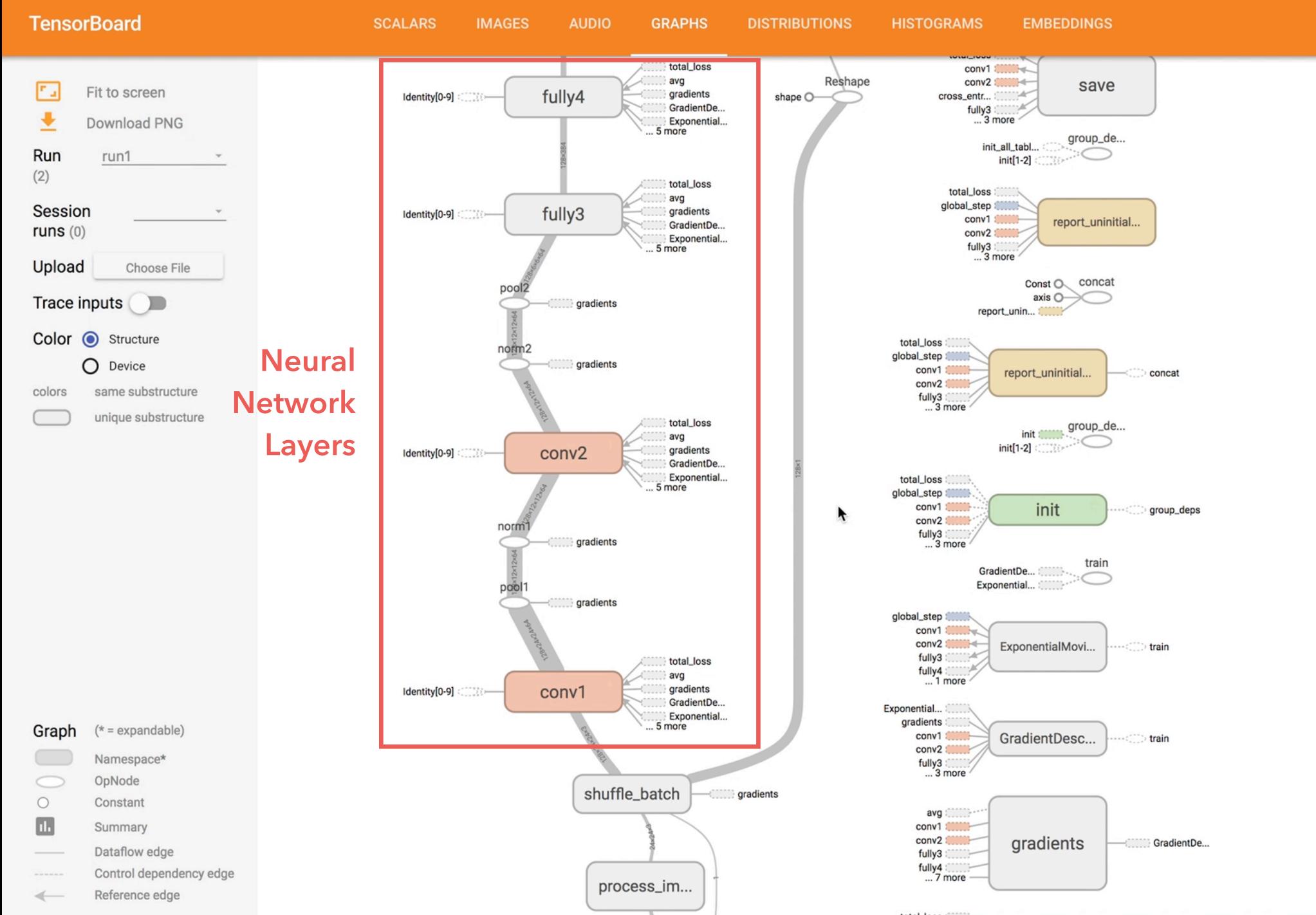


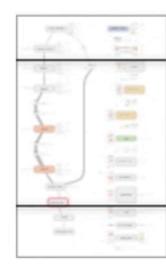






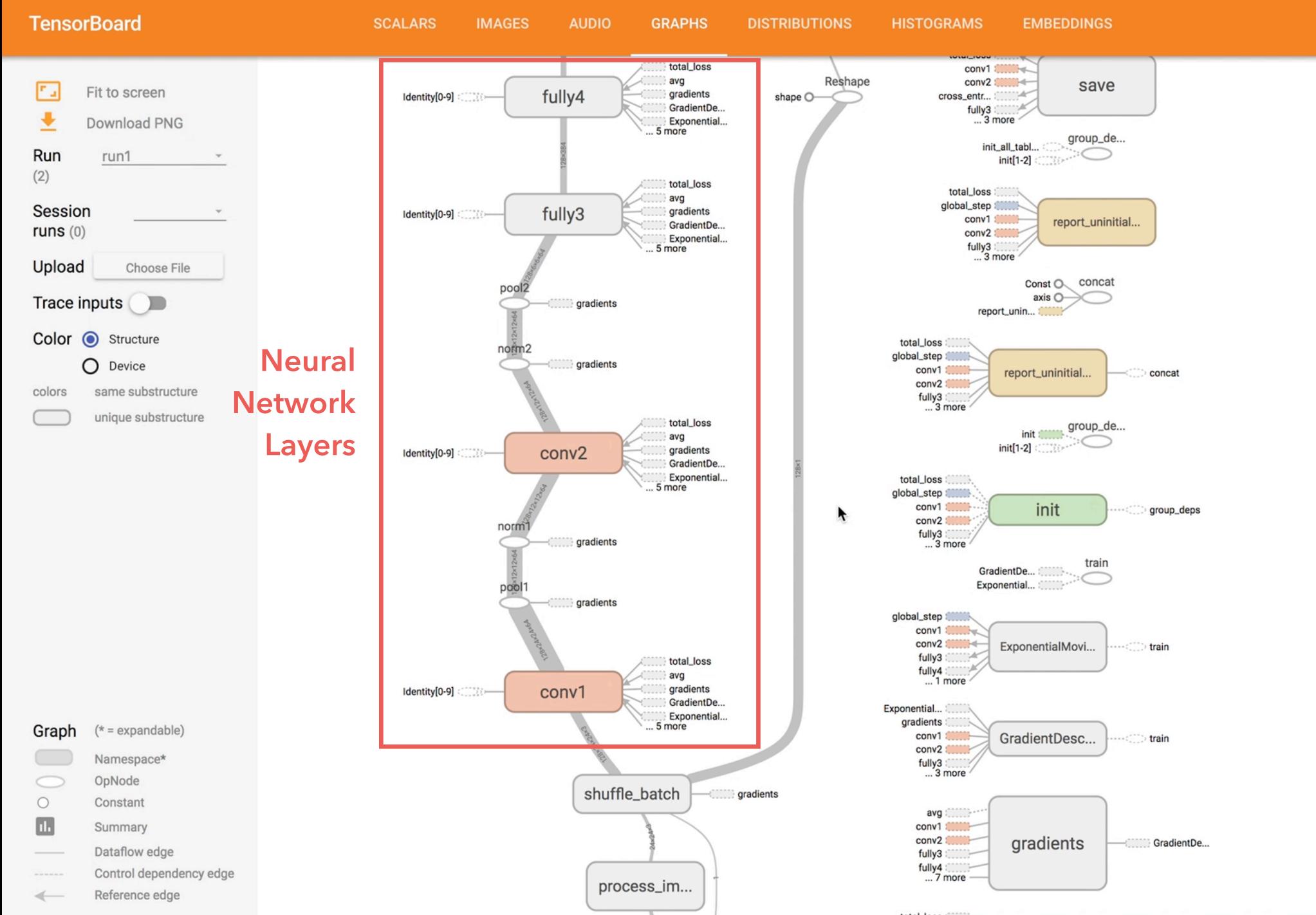


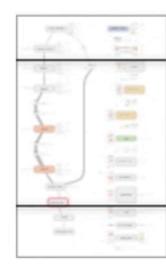




?

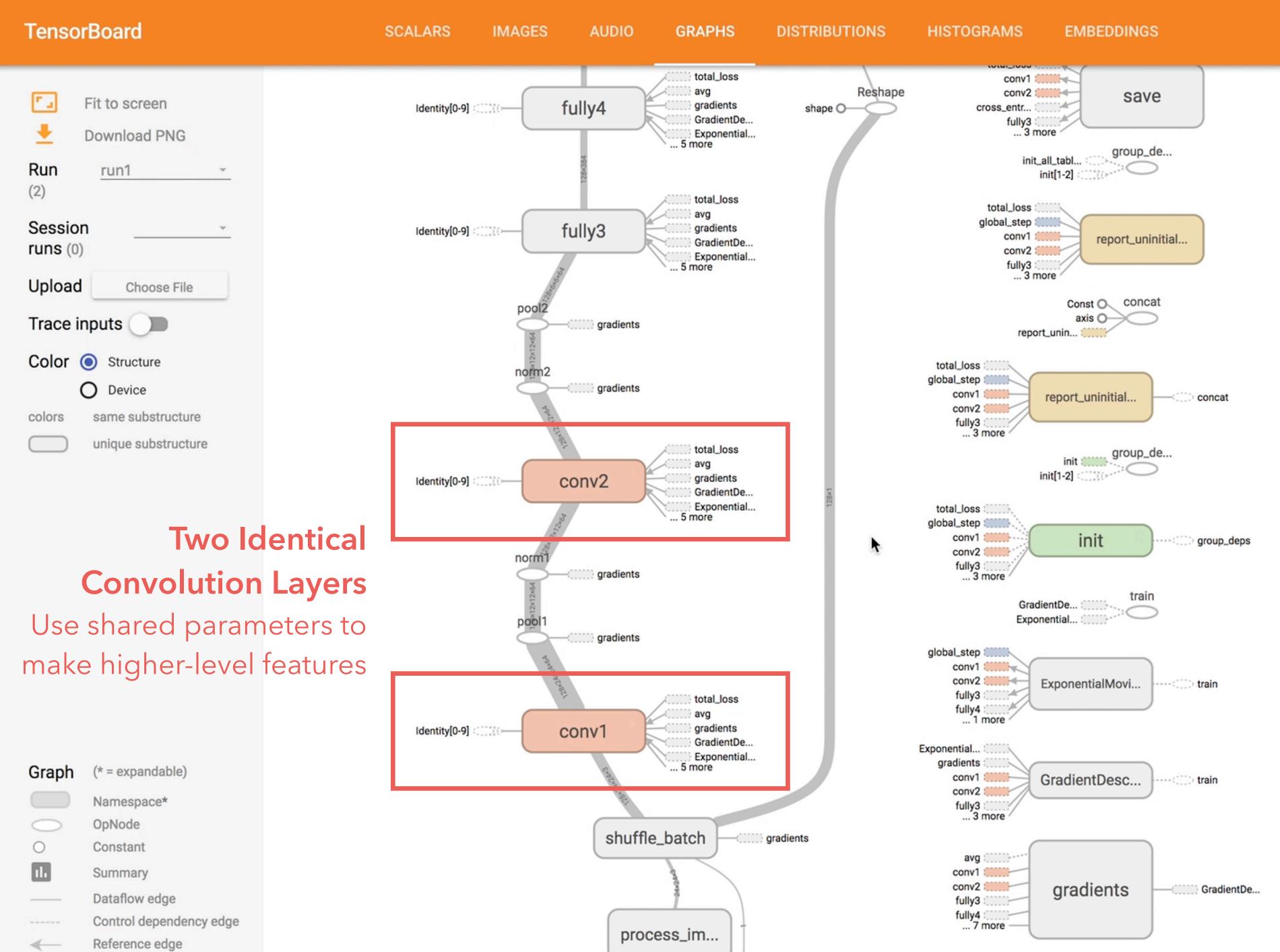






?

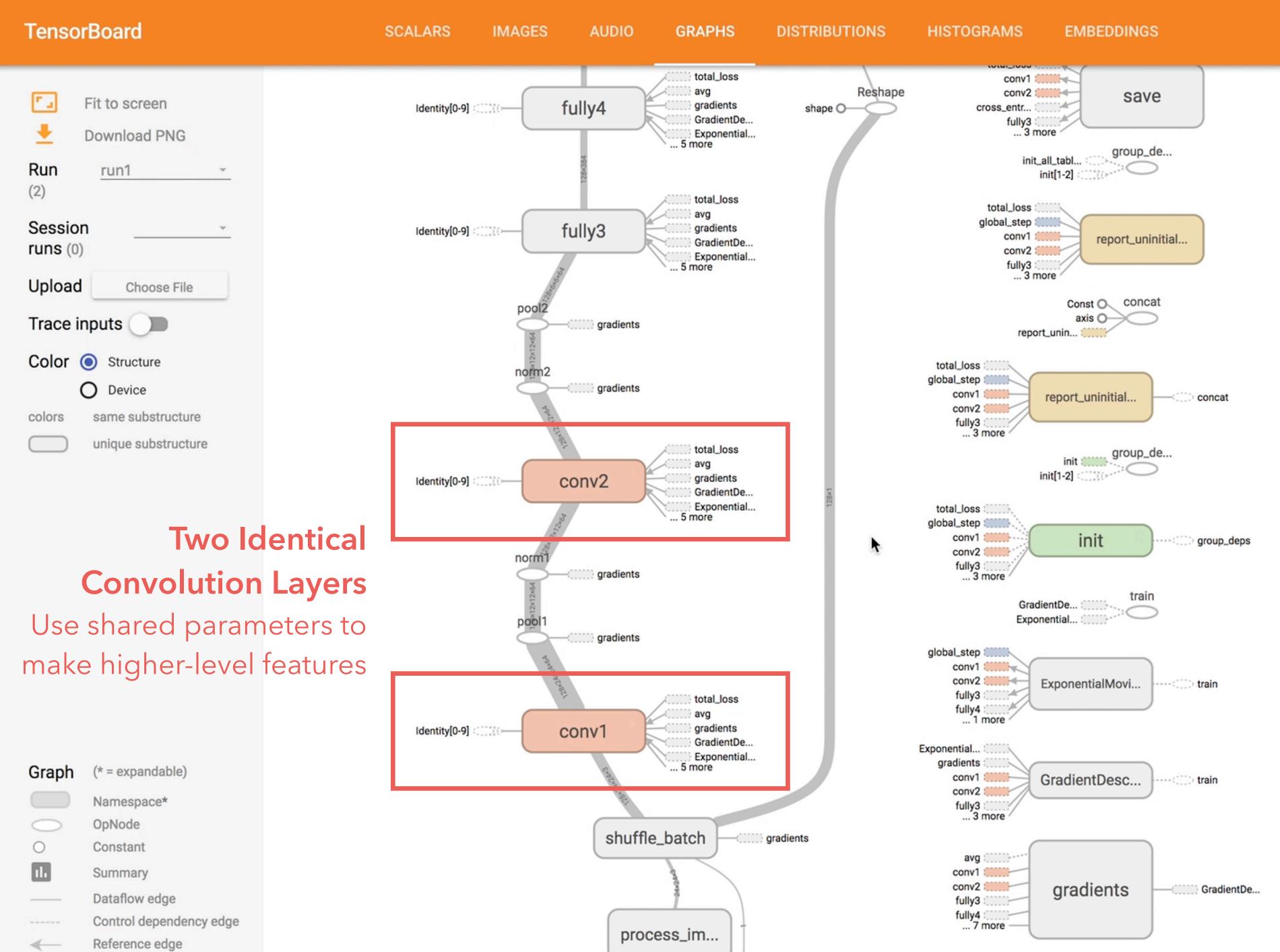






?

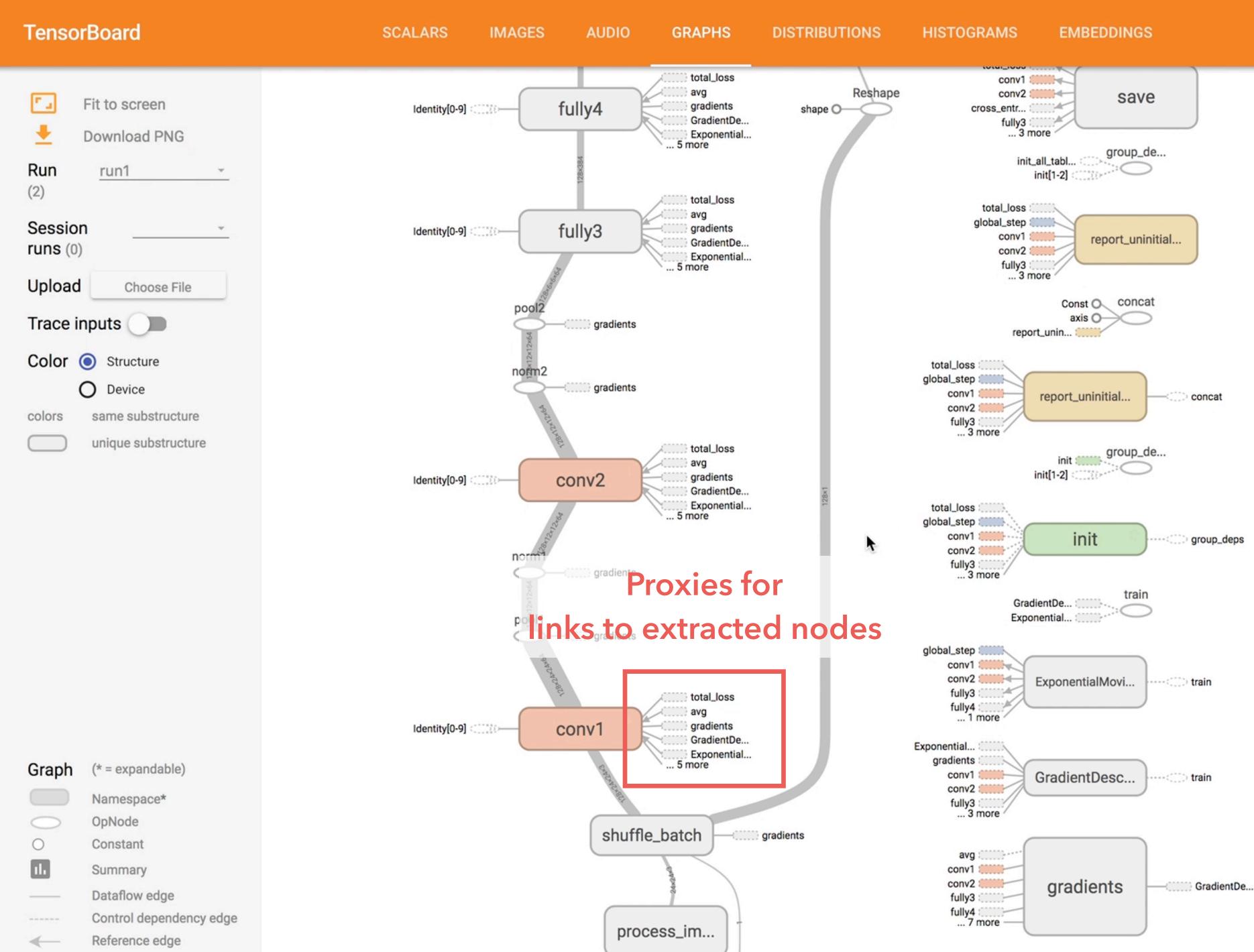


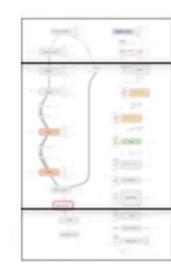




?

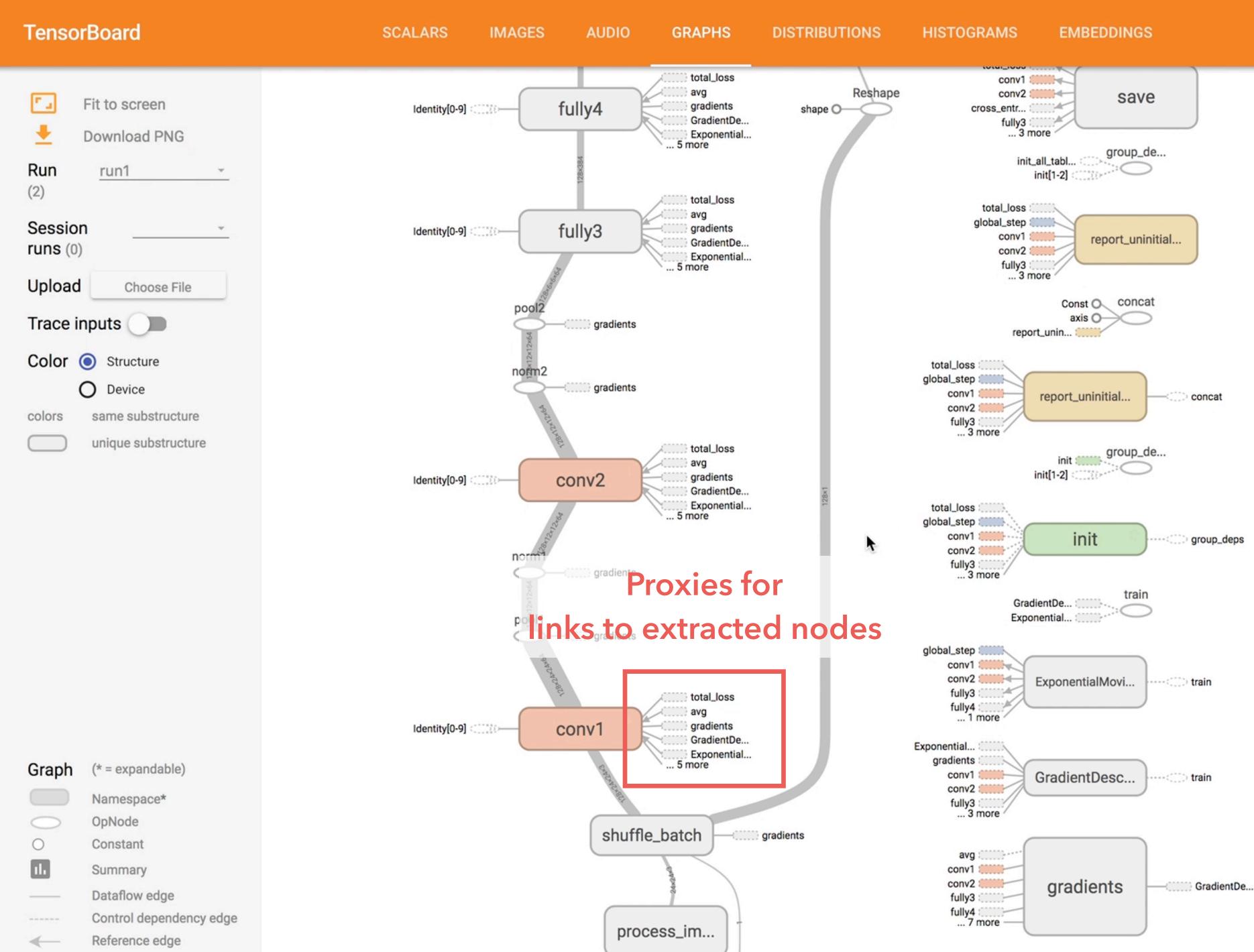


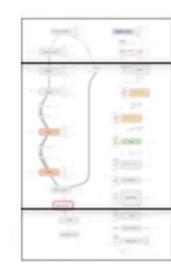




?

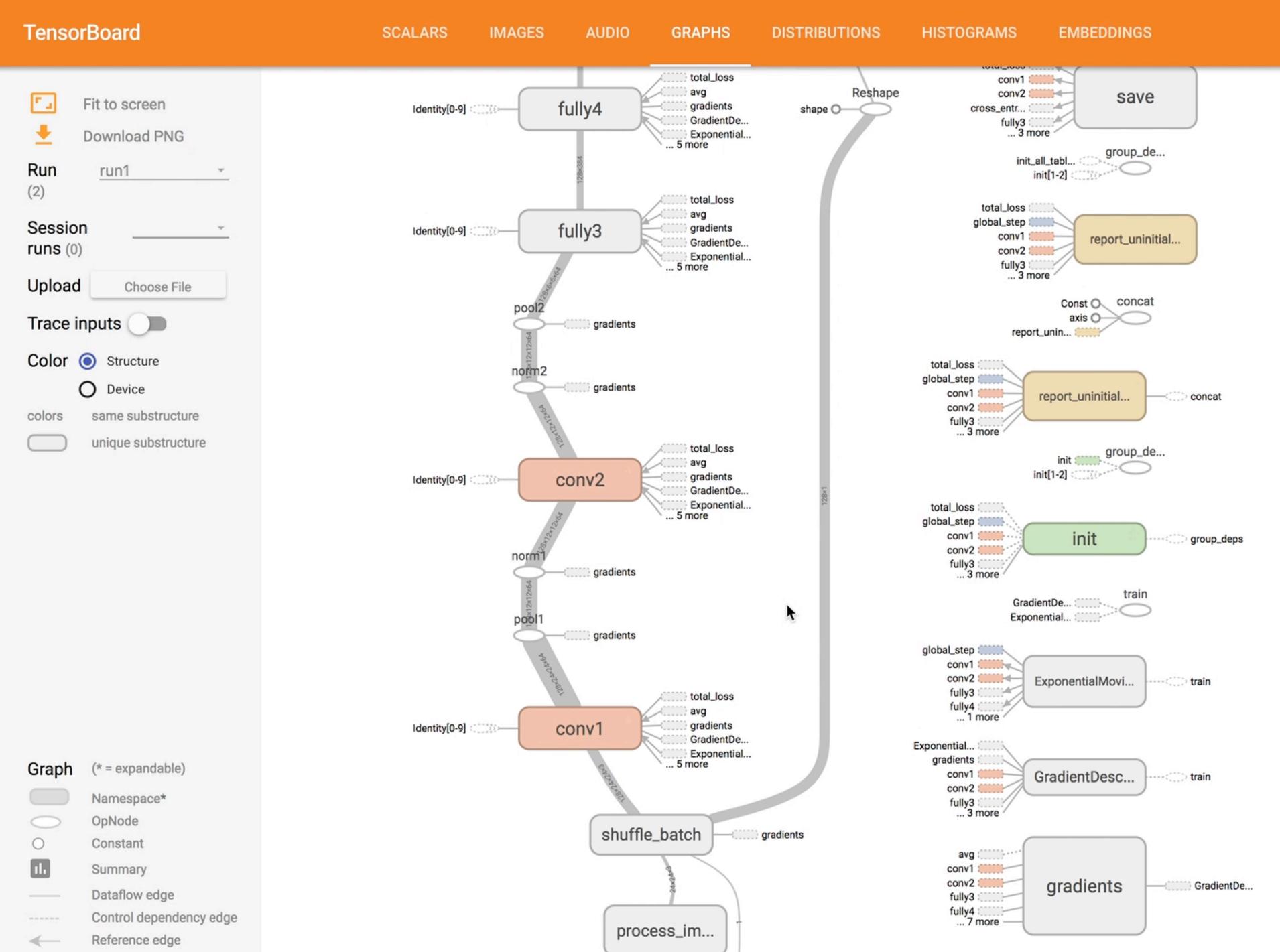


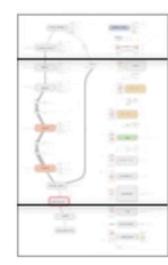




?

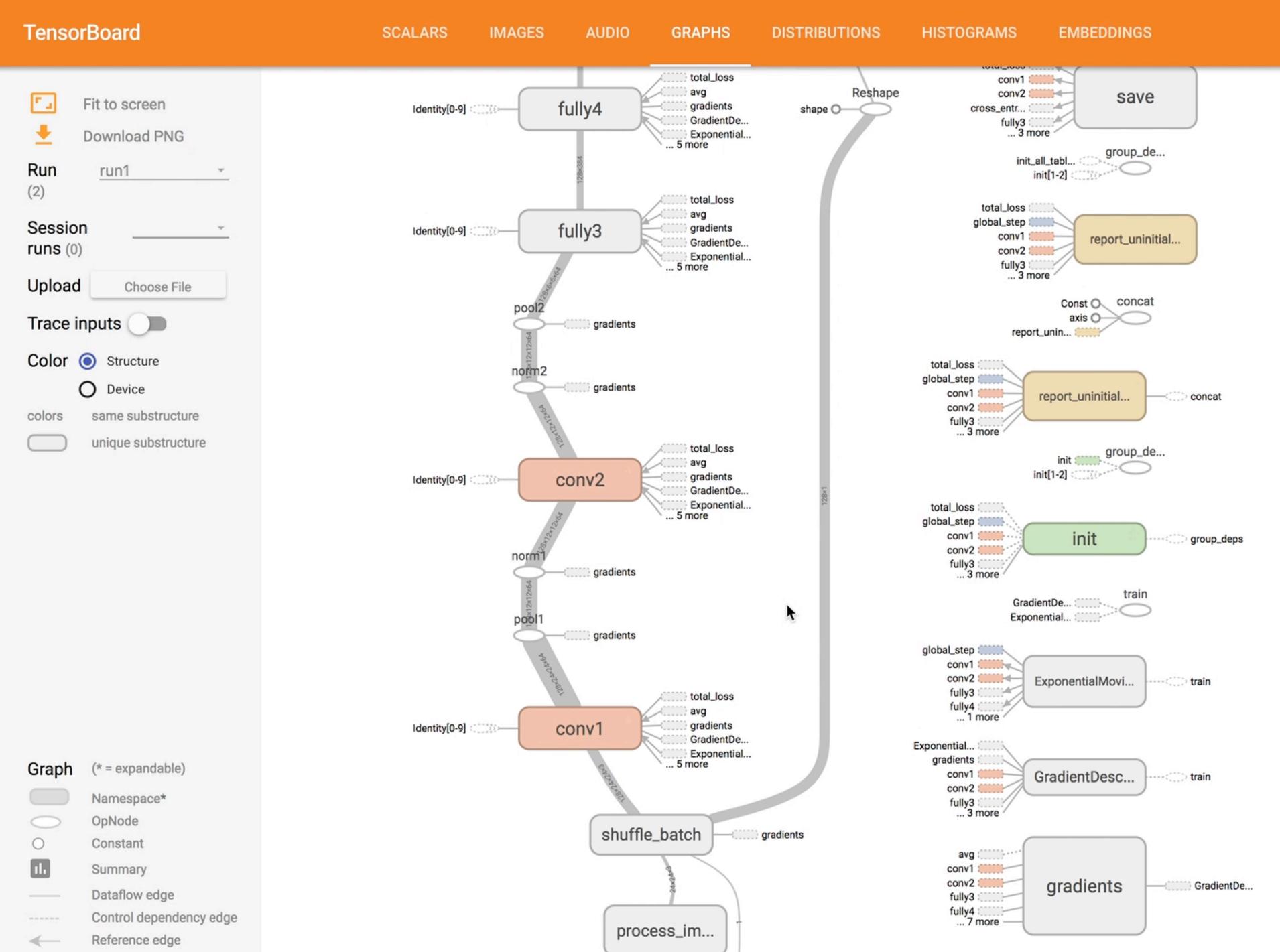


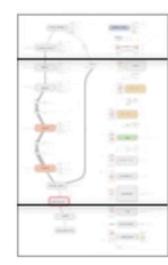




?

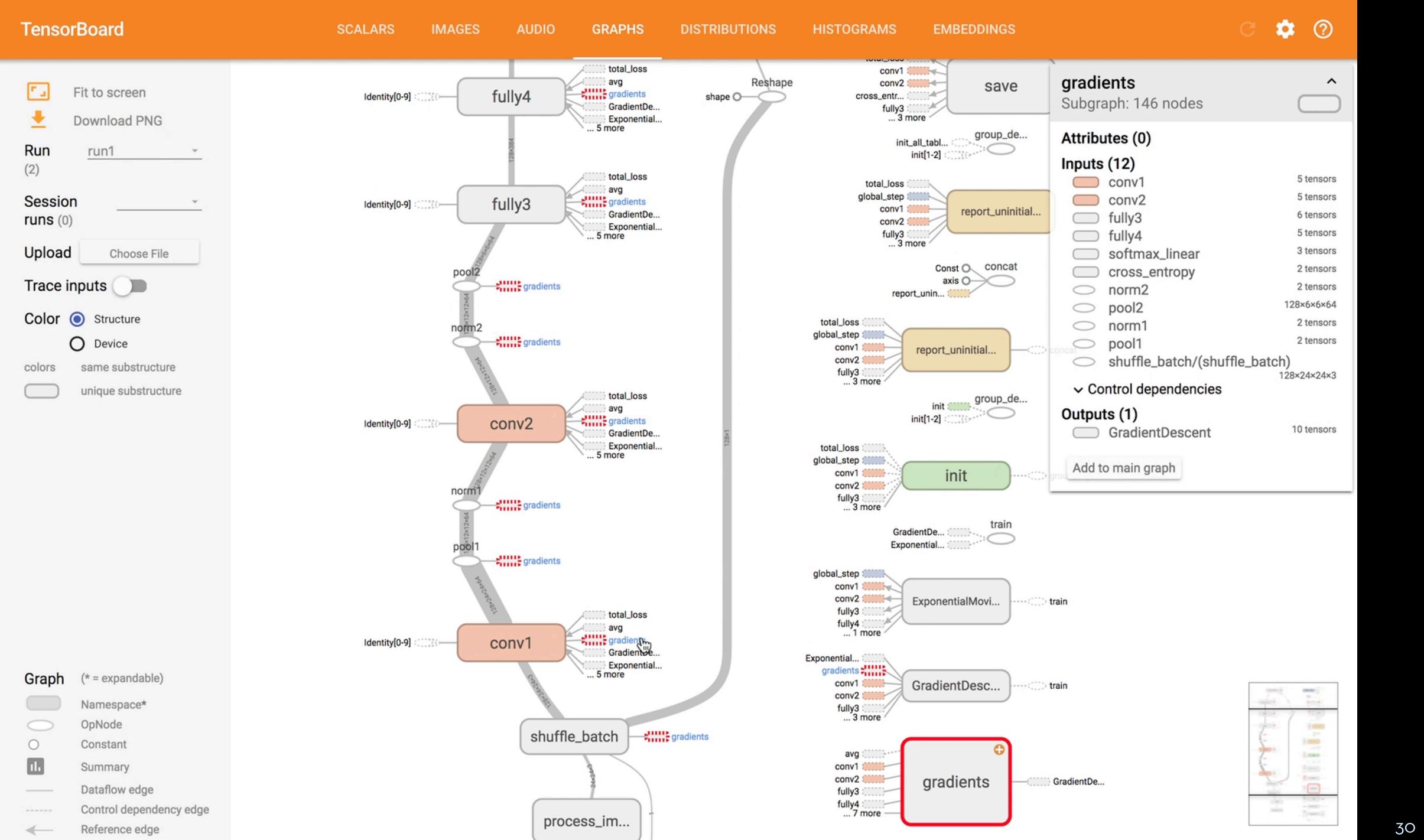


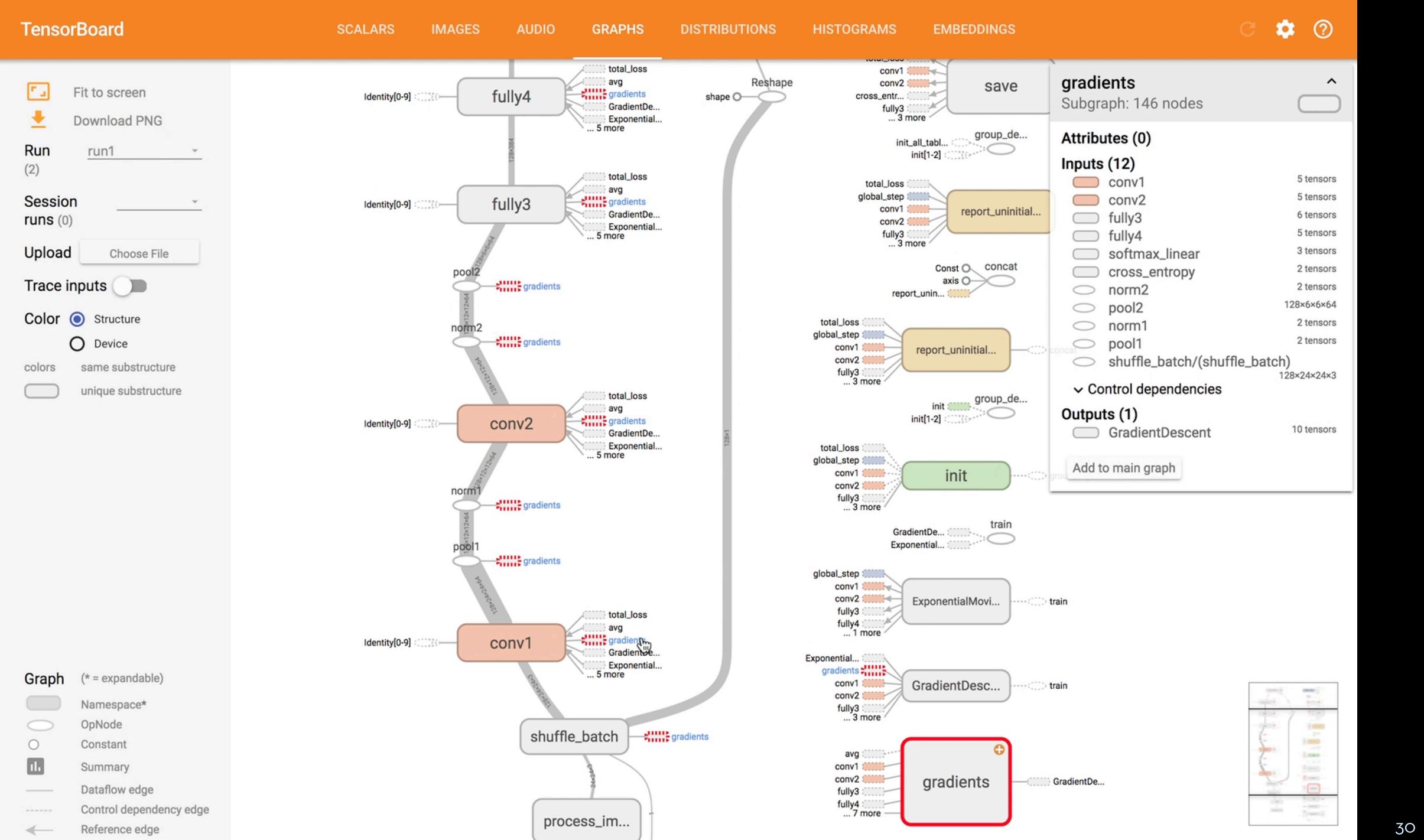


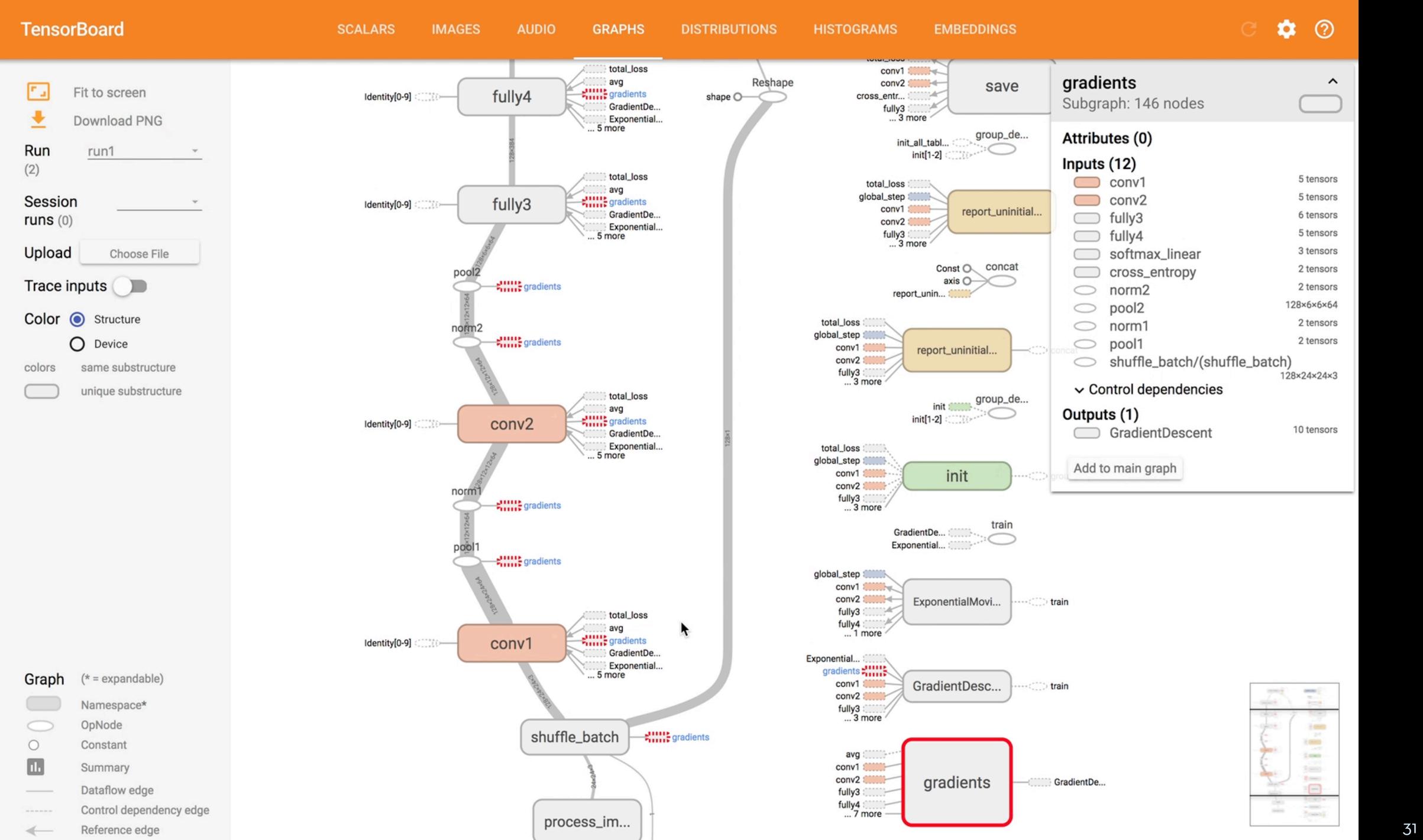


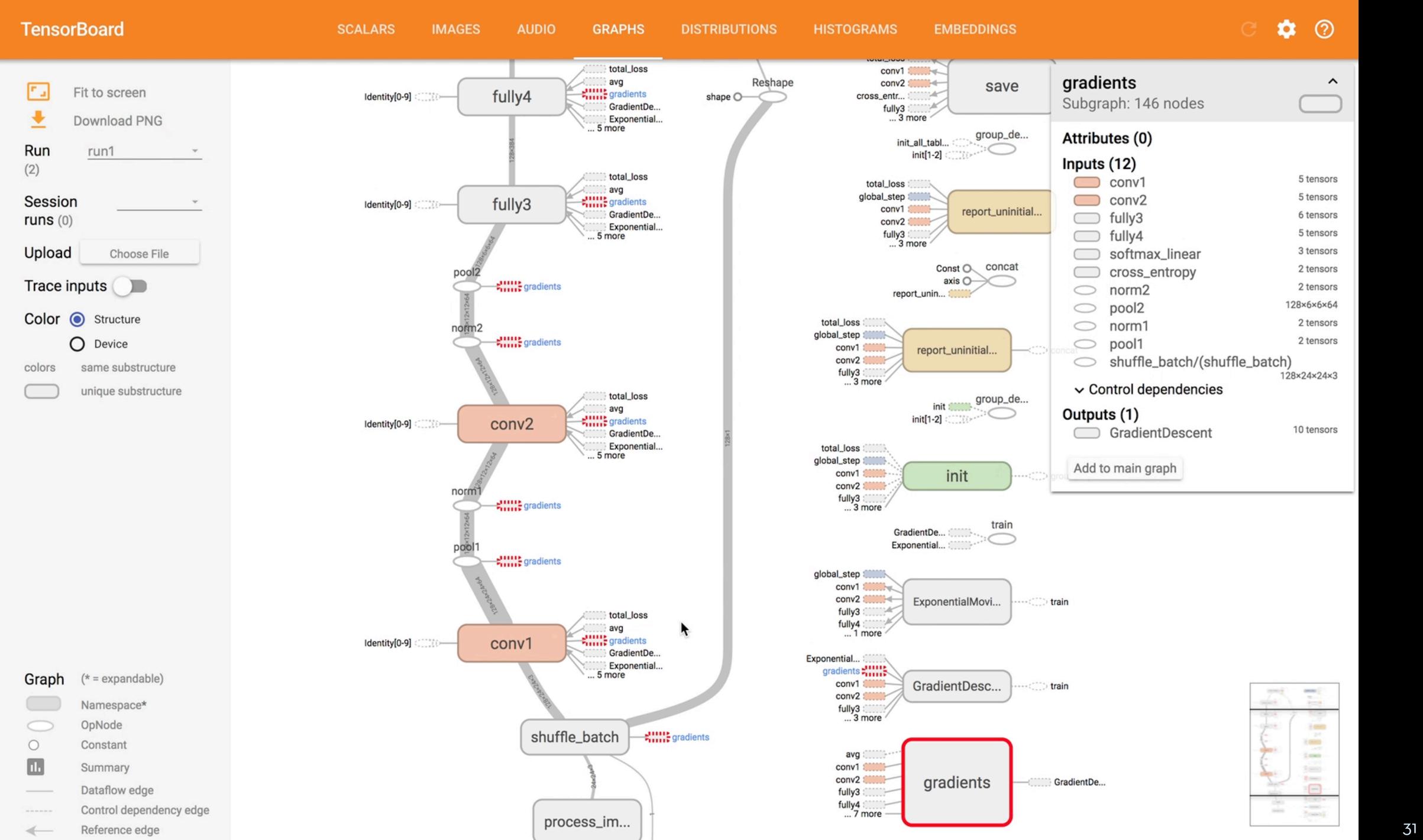
?

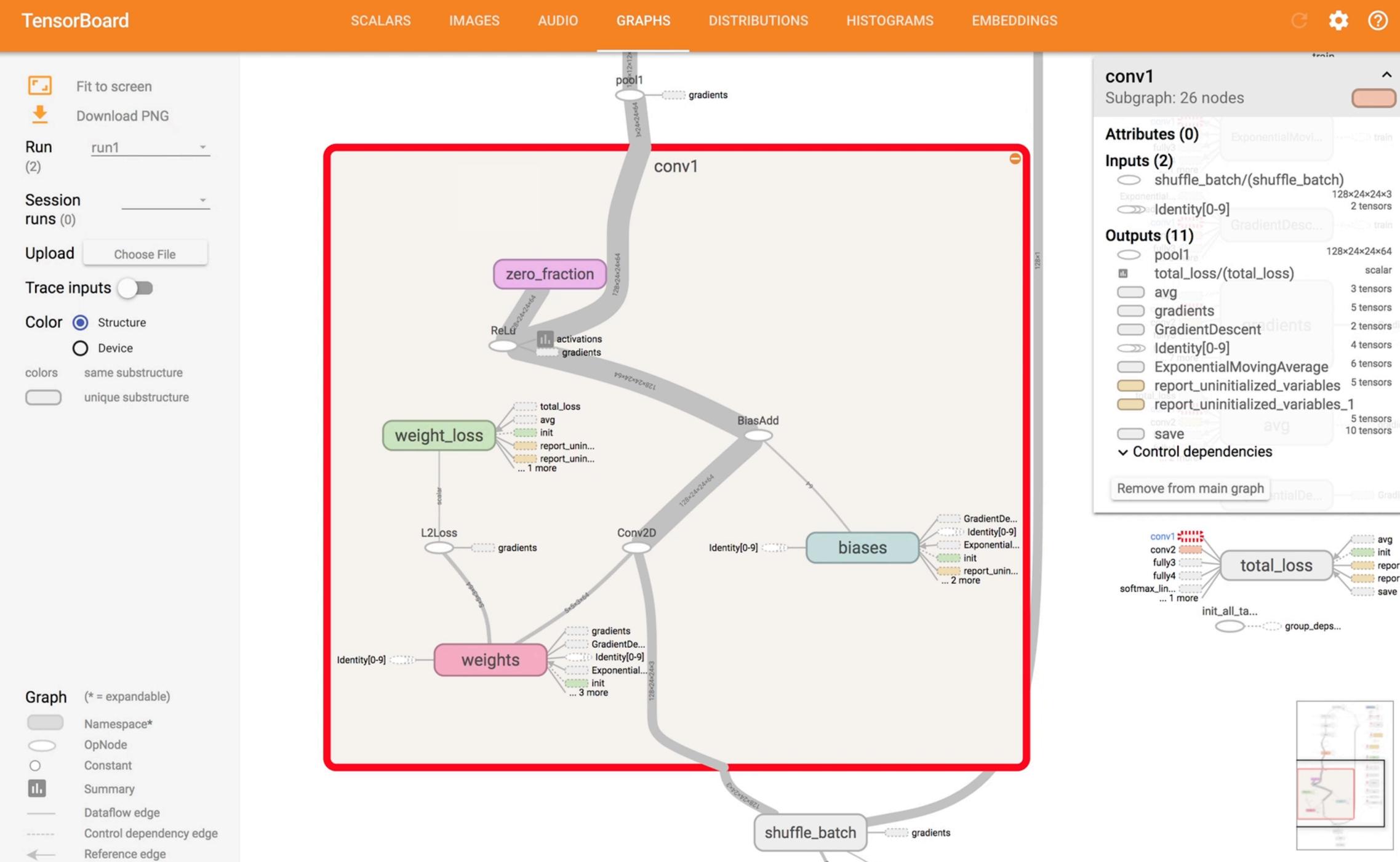


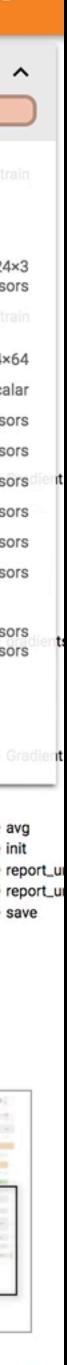


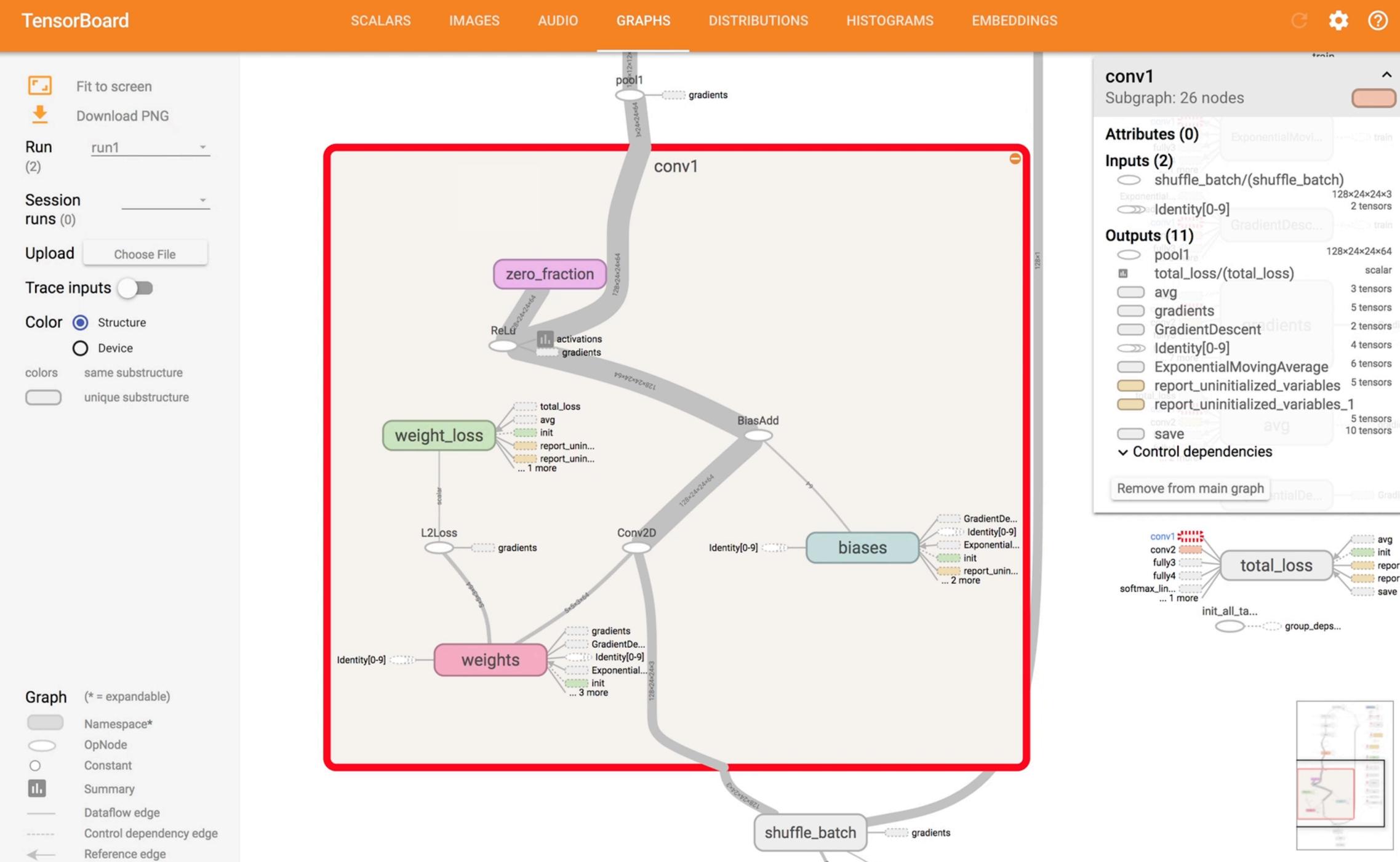


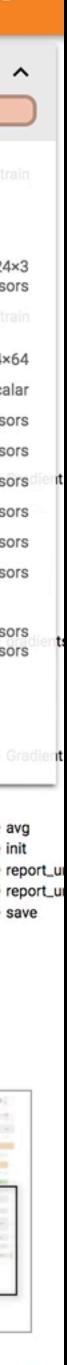


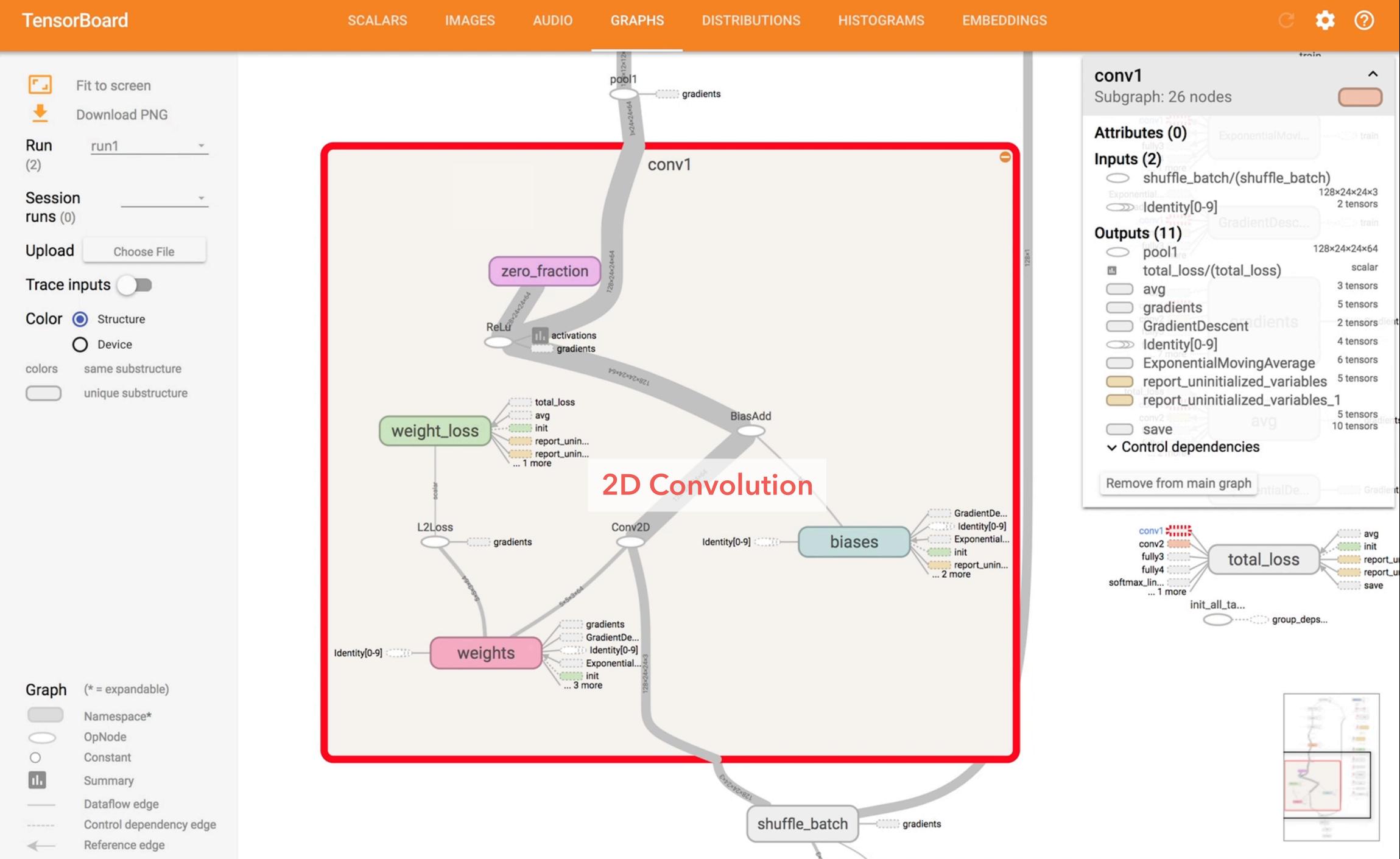


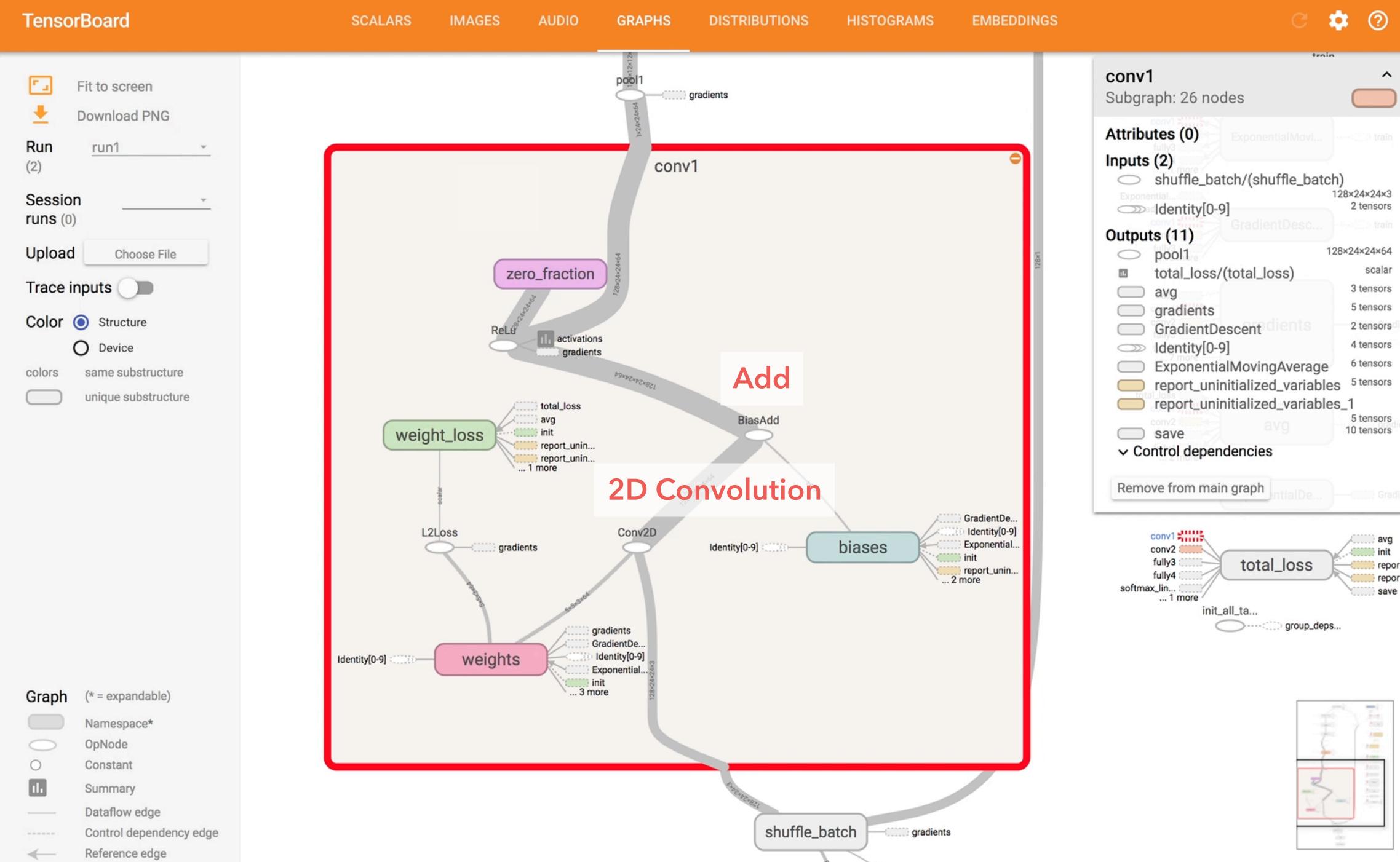






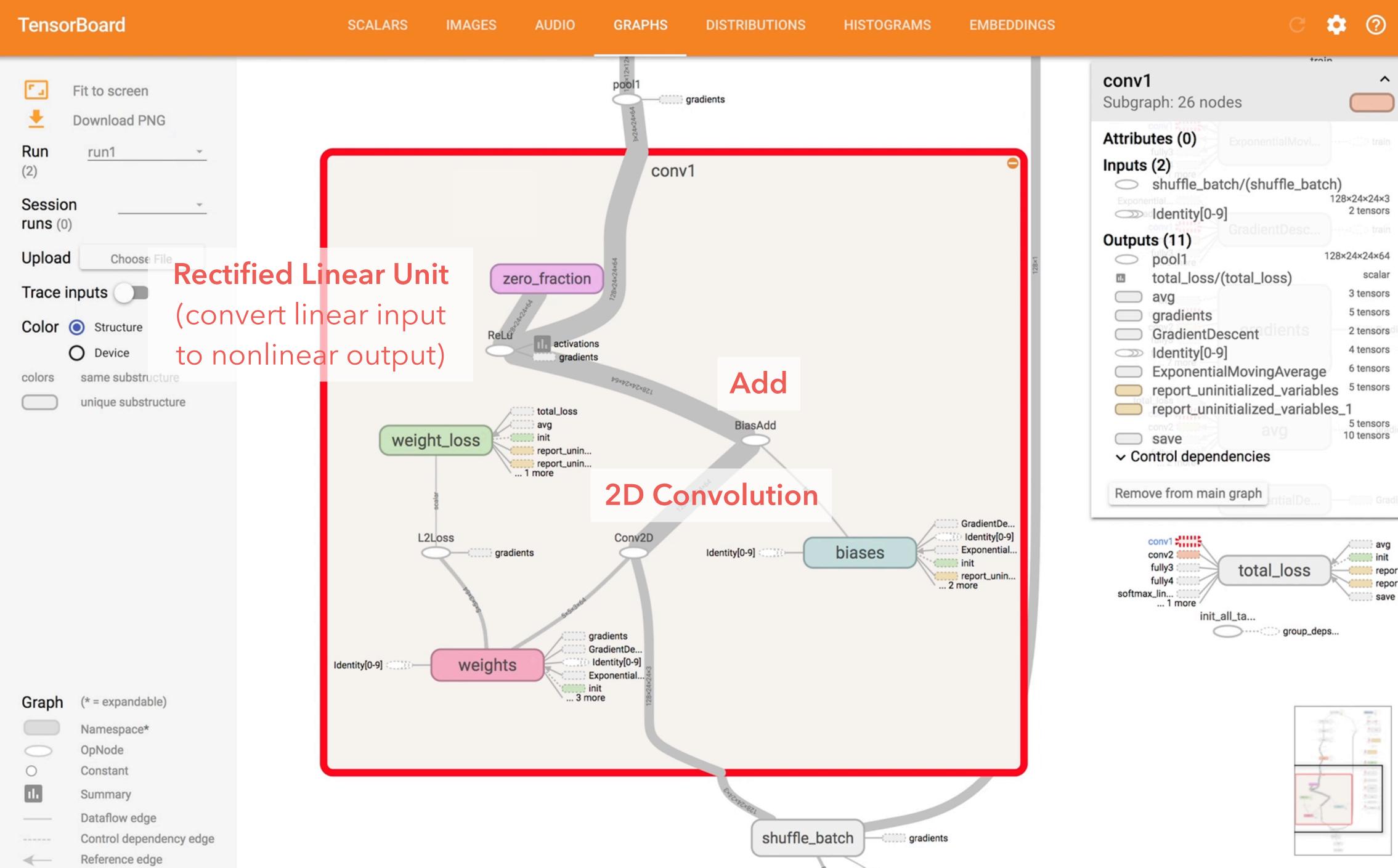






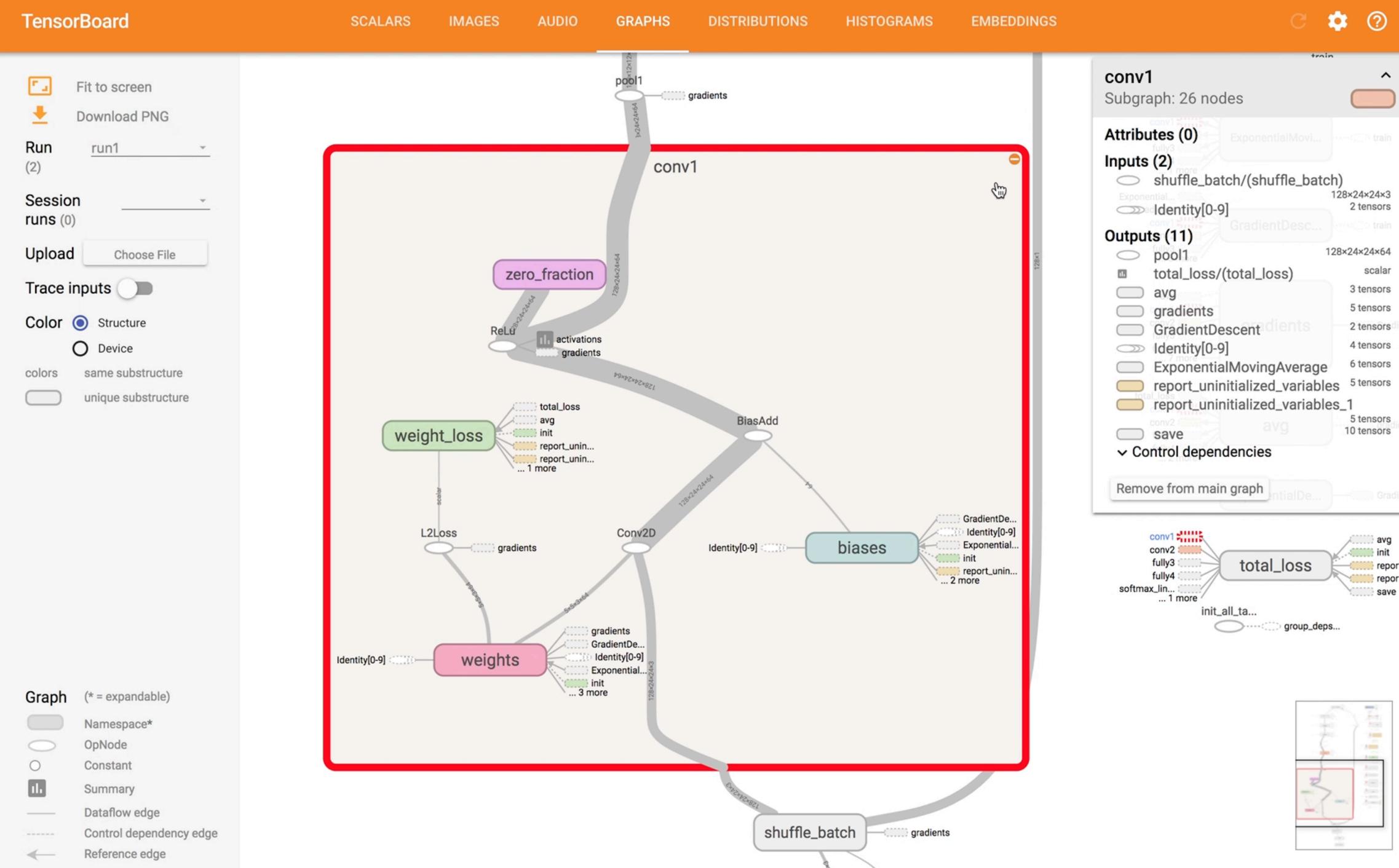




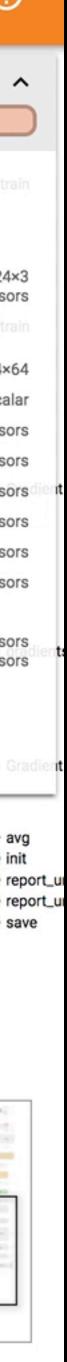


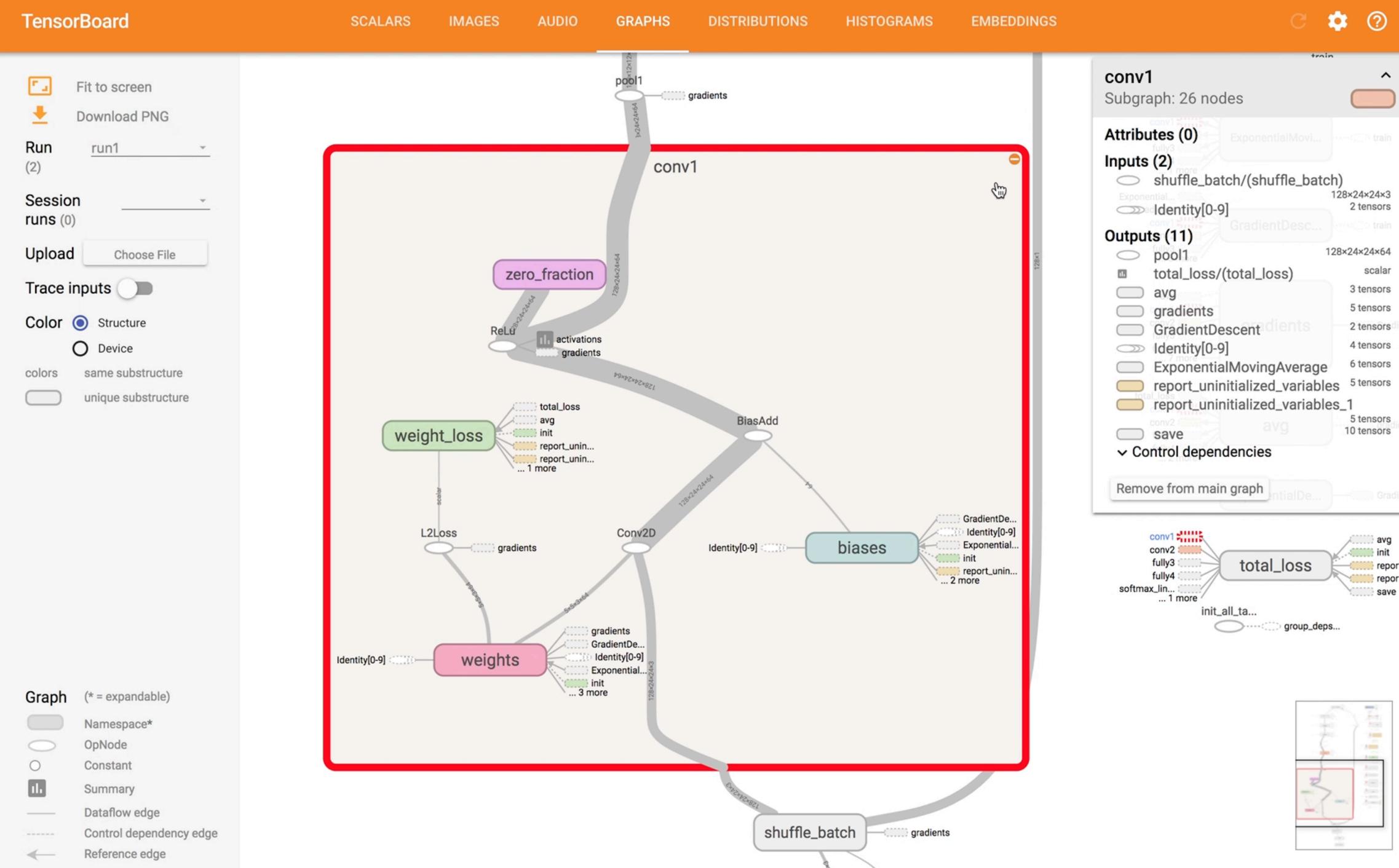




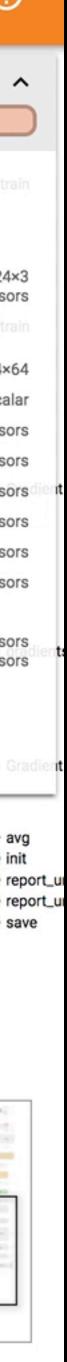


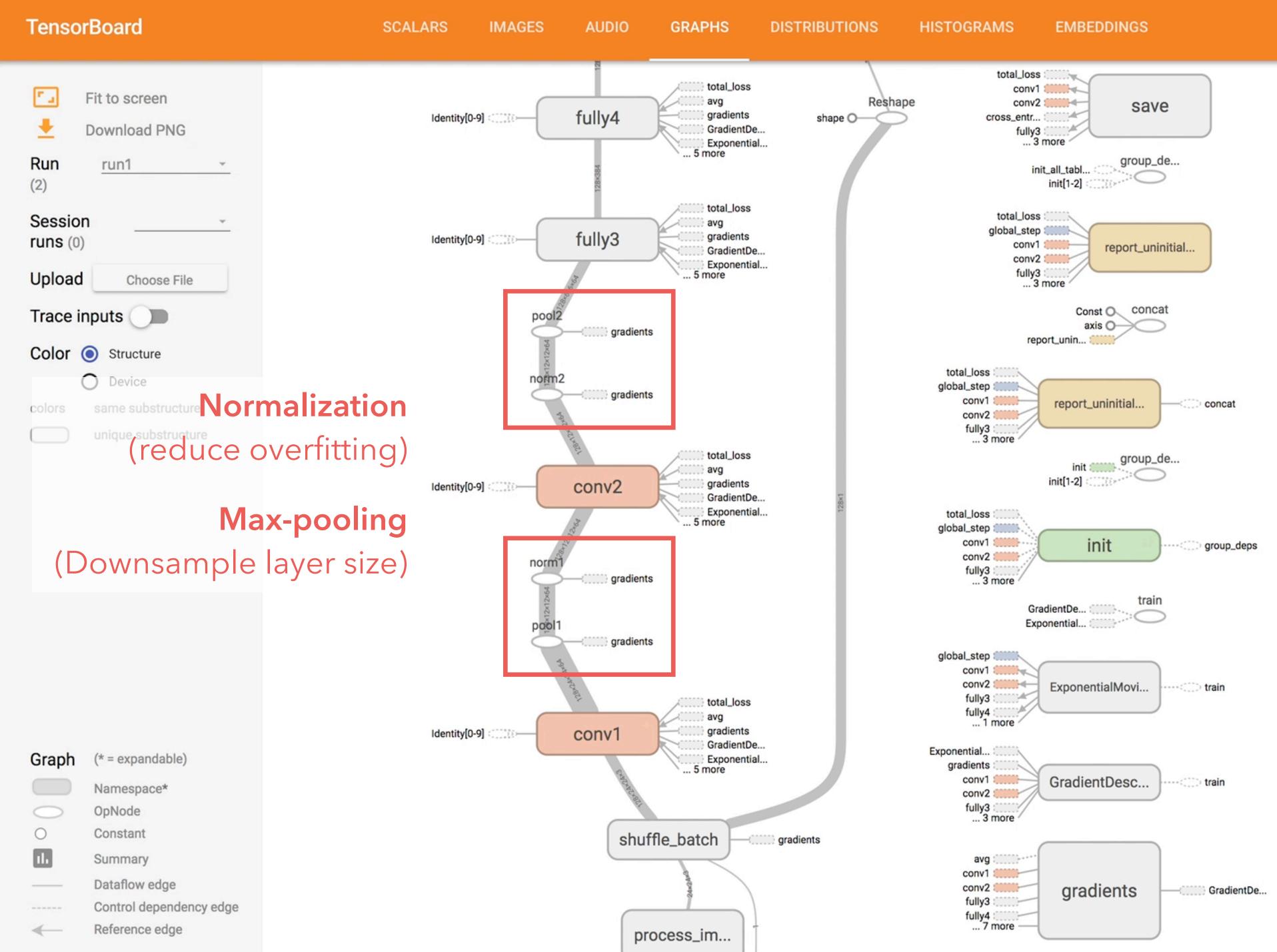


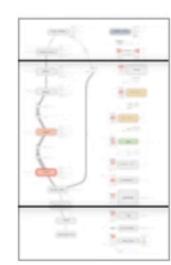






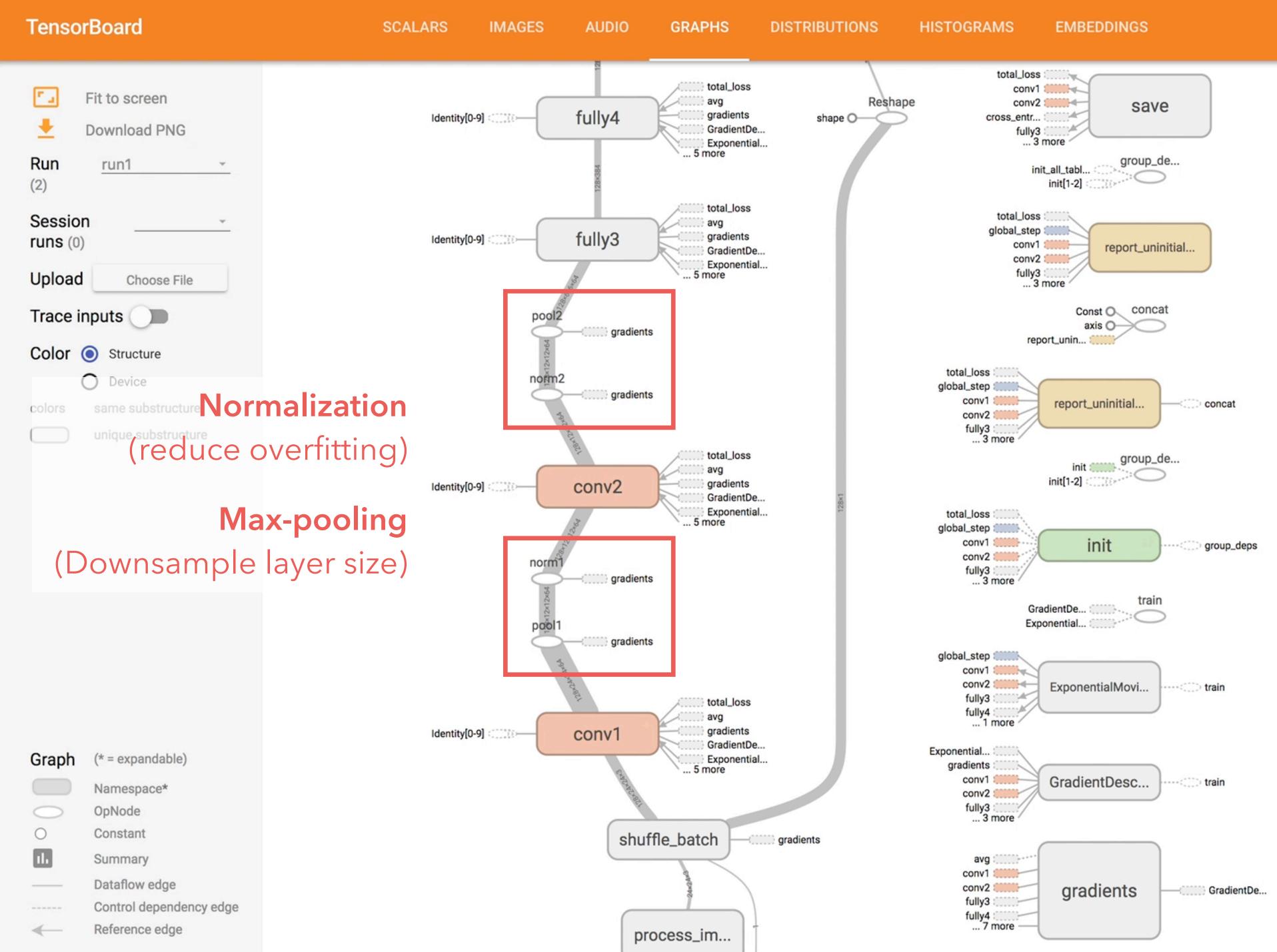


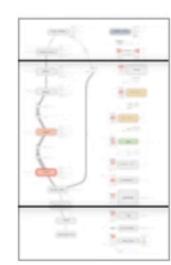




?







?

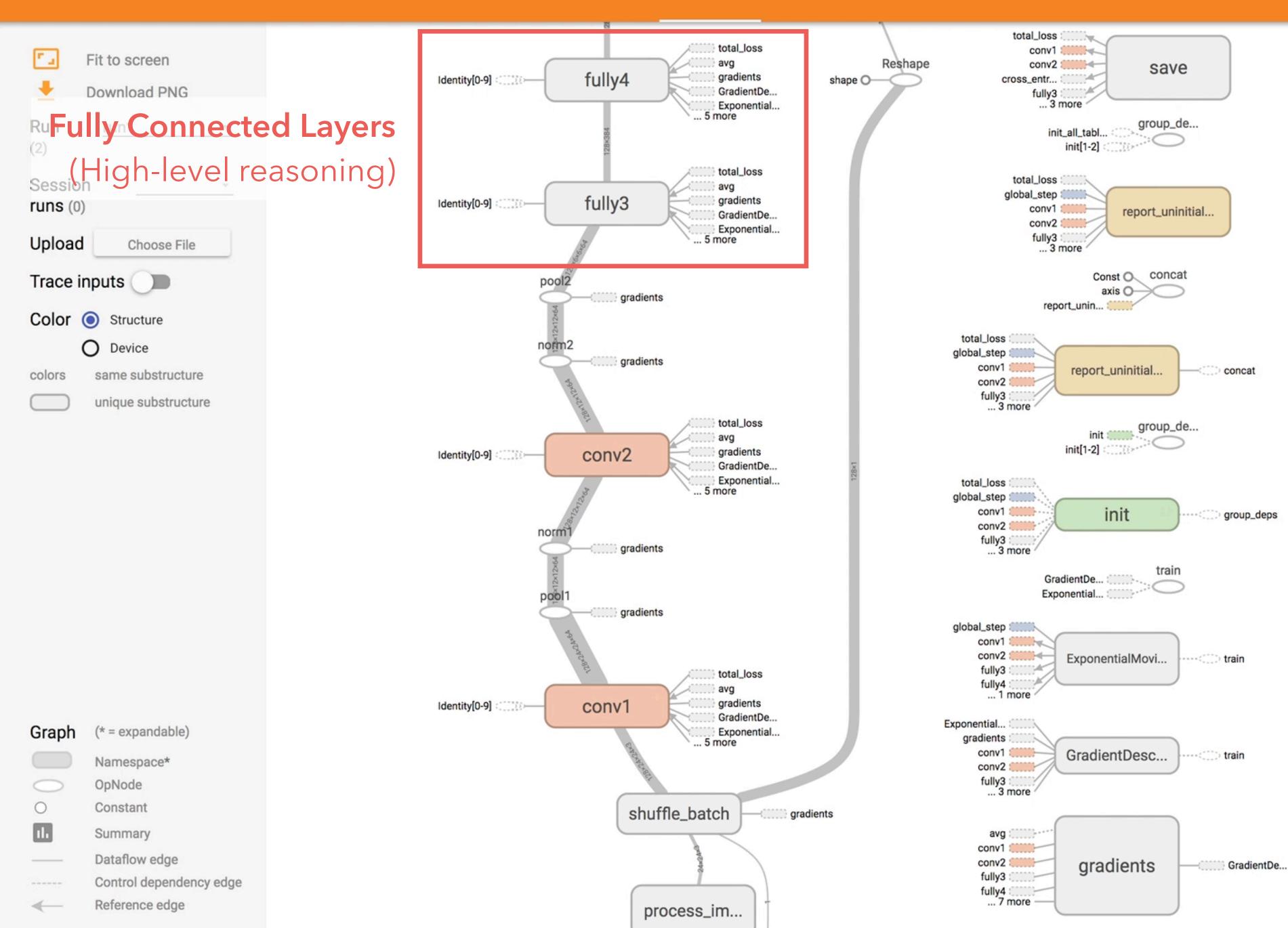


TensorBoard

IMAGES

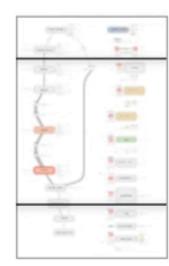
AUDIO

GRAPHS



IS DISTRIBUTIONS HISTOGRAMS EMBEDDINGS





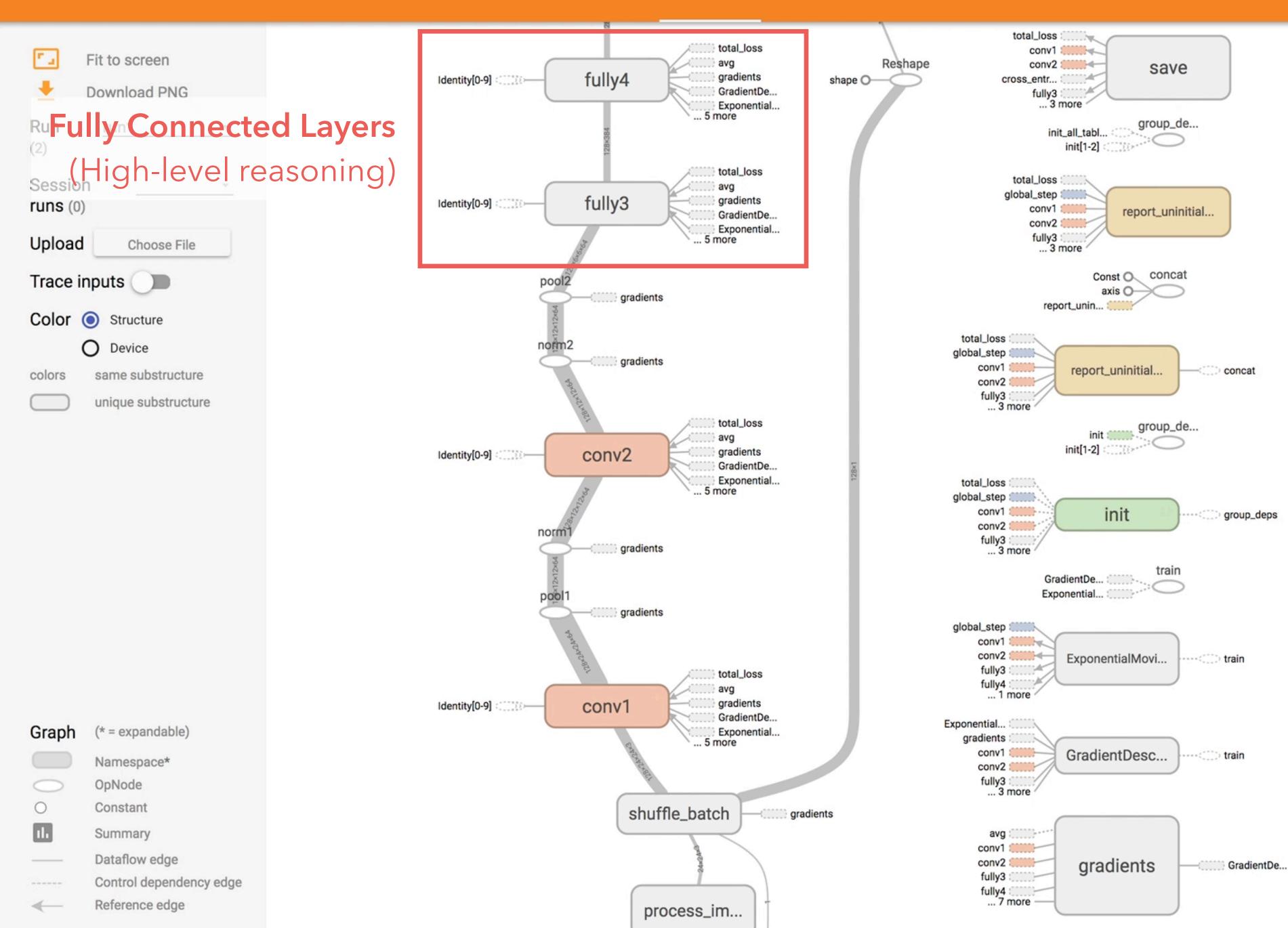


TensorBoard

IMAGES

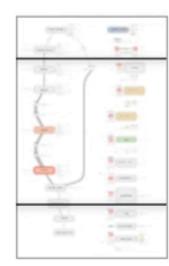
AUDIO

GRAPHS

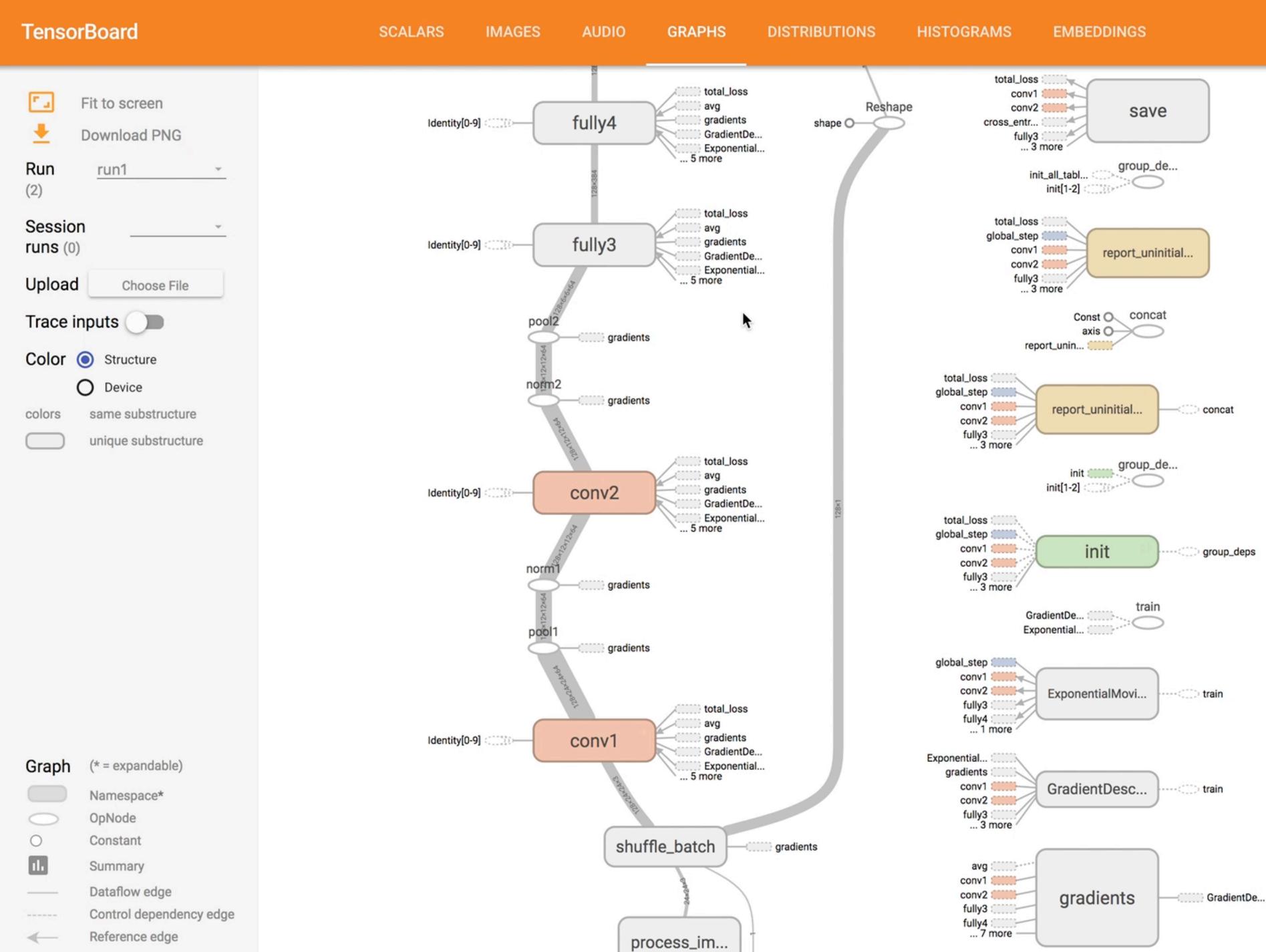


IS DISTRIBUTIONS HISTOGRAMS EMBEDDINGS







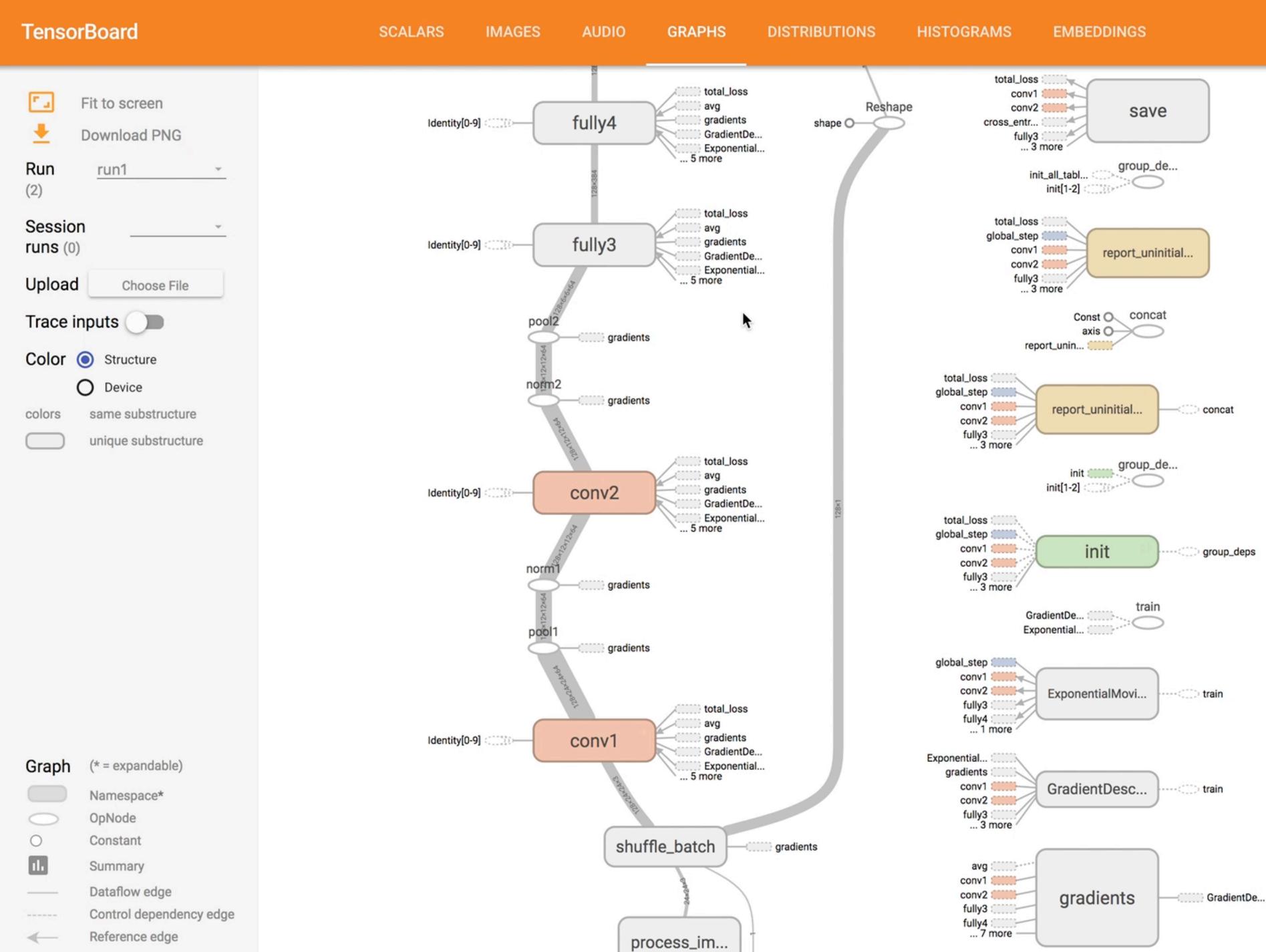




?





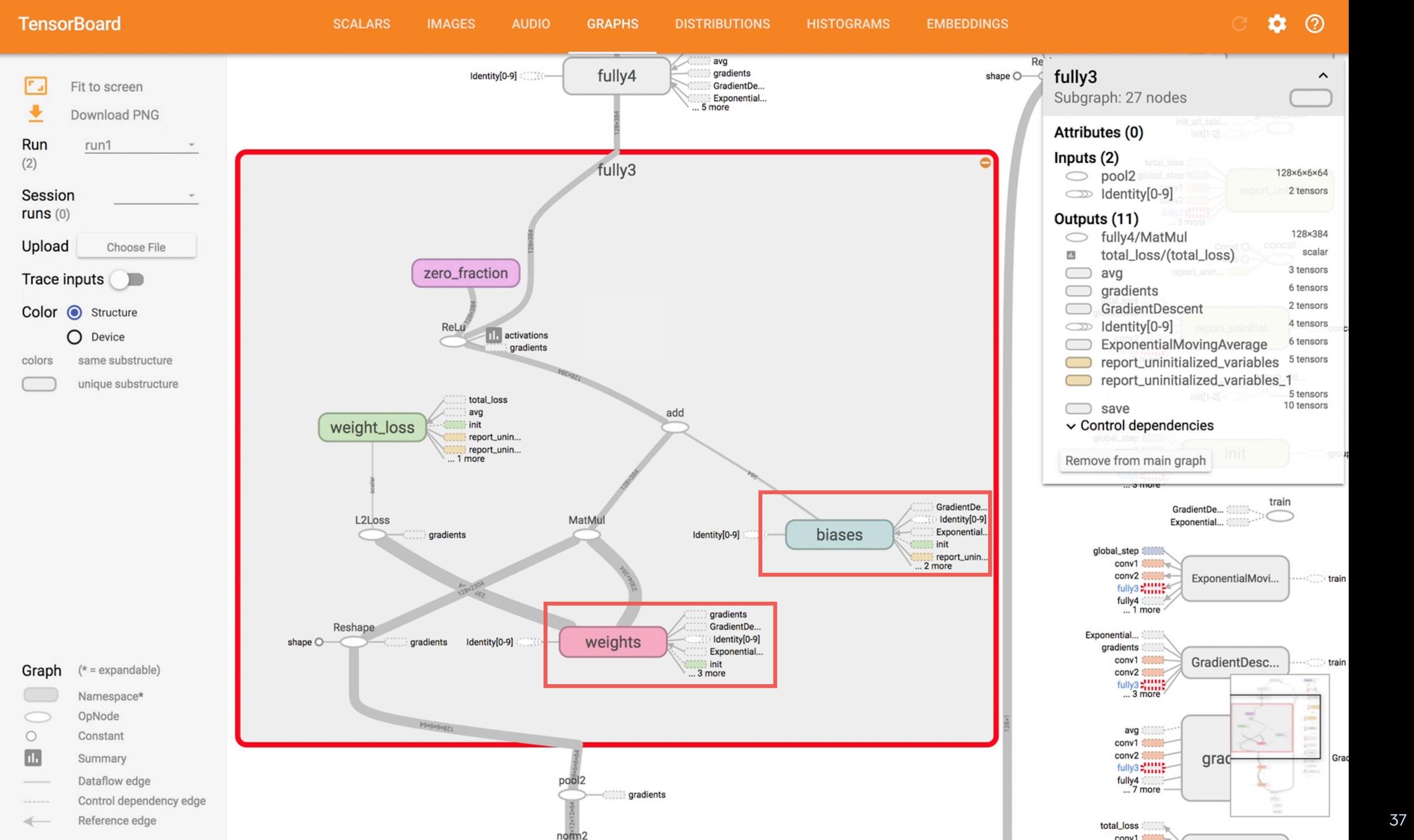


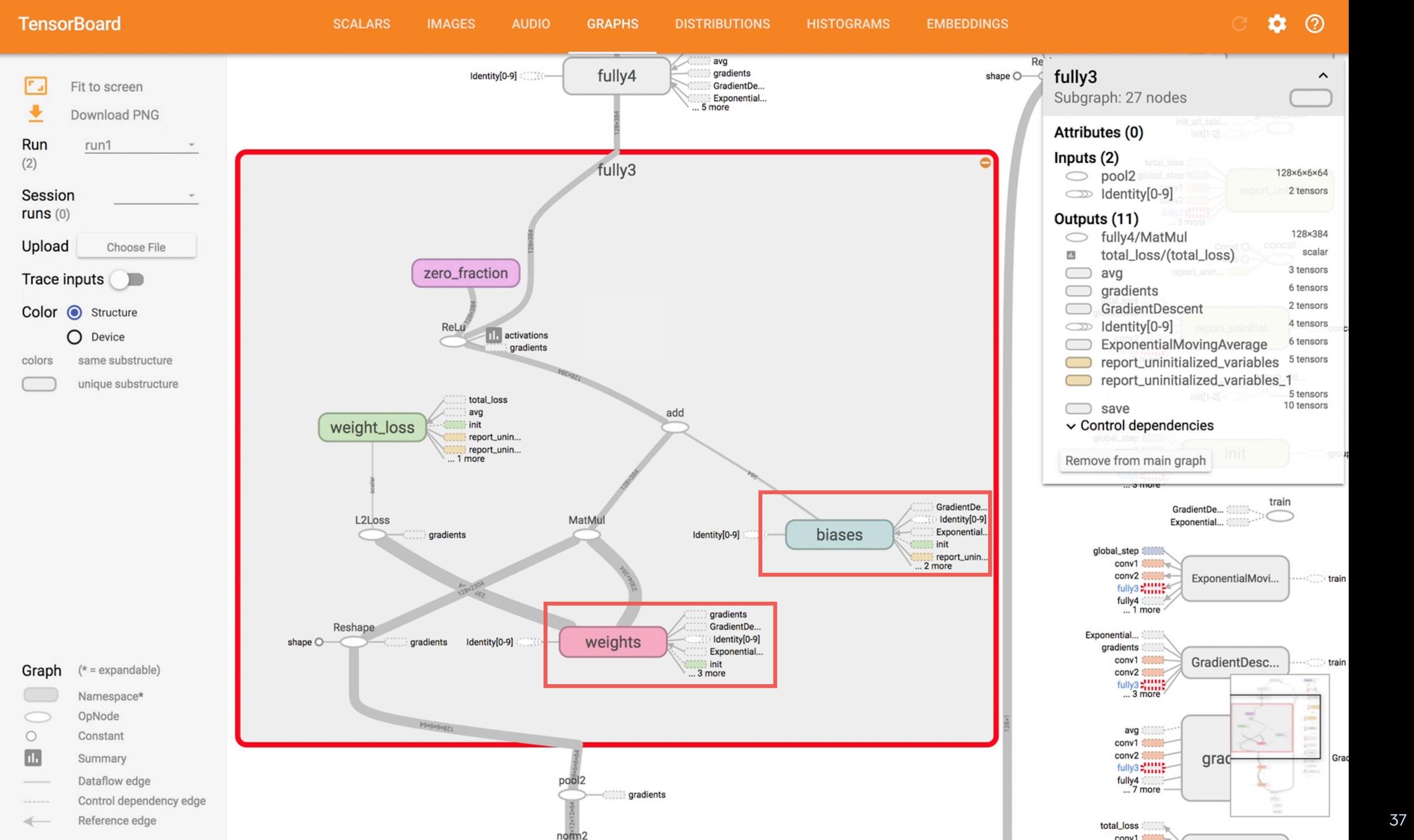


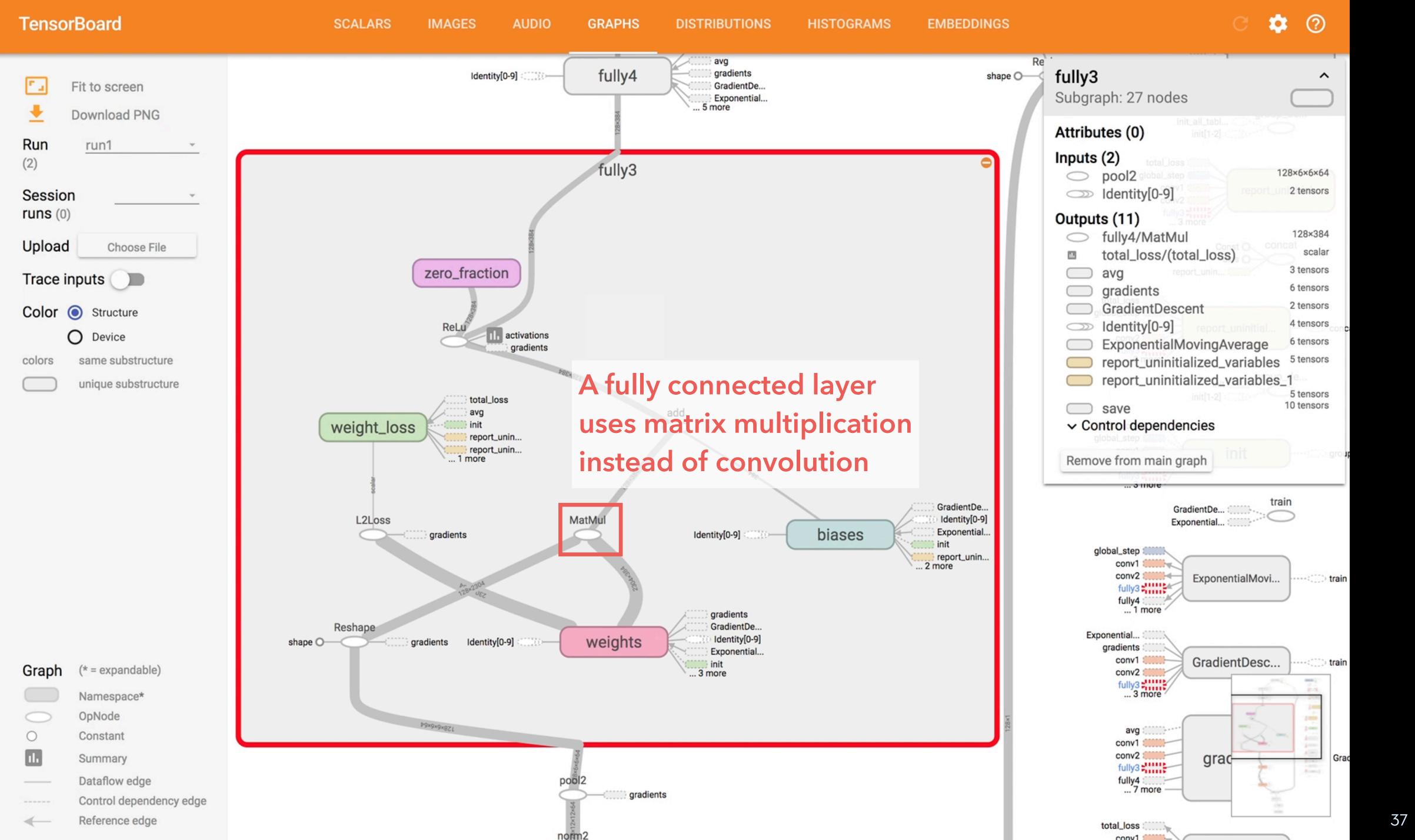
?

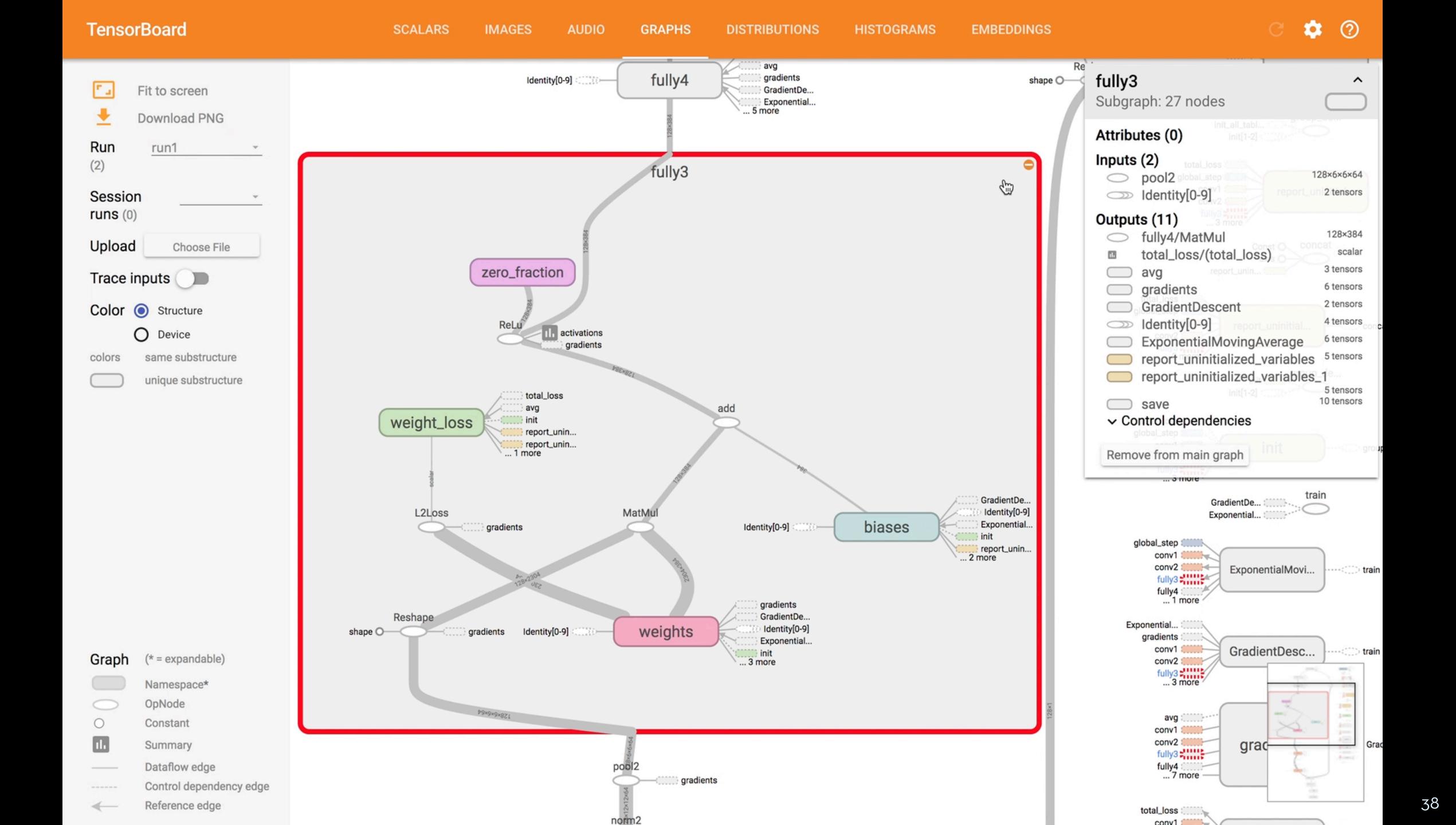


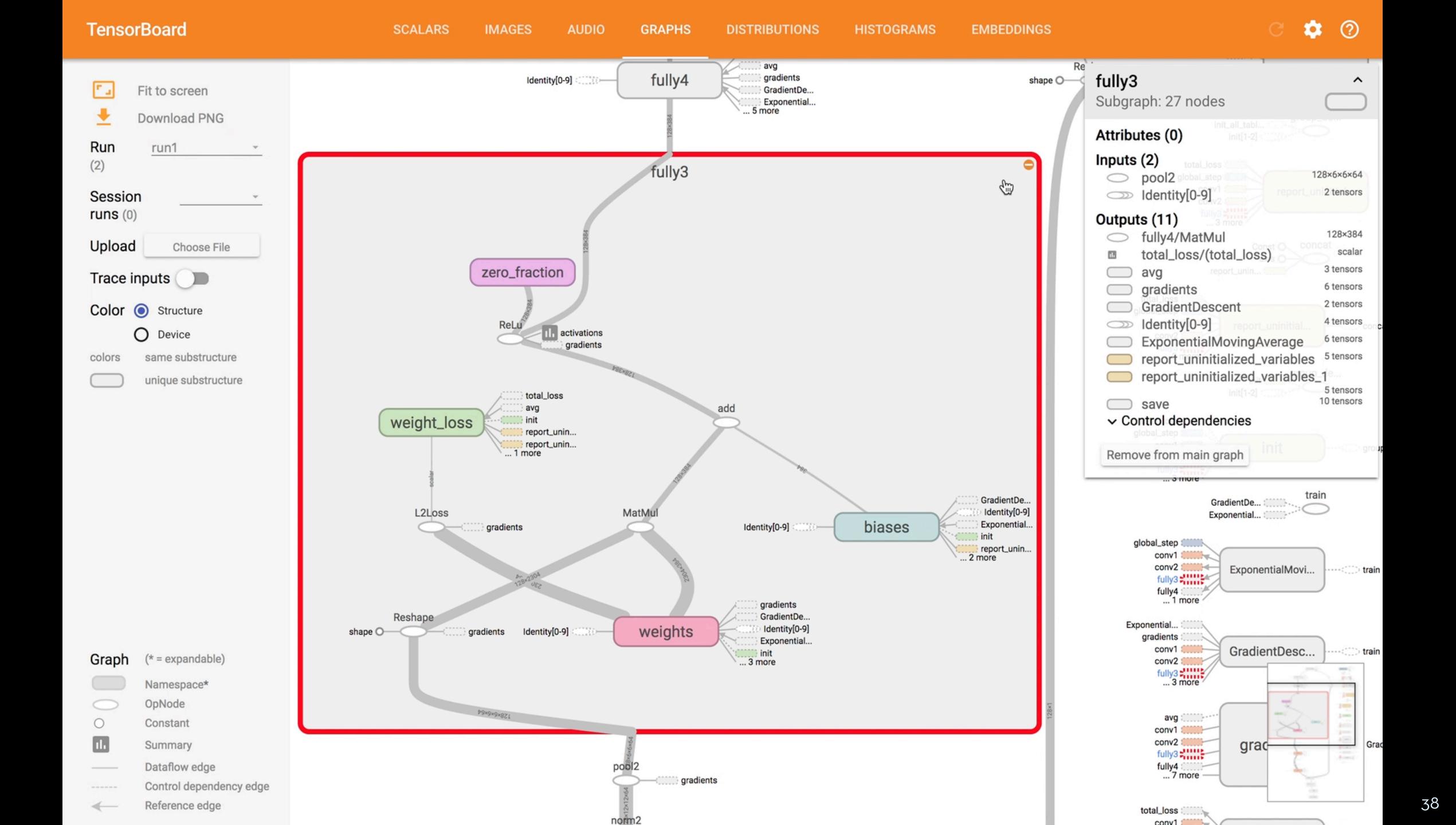






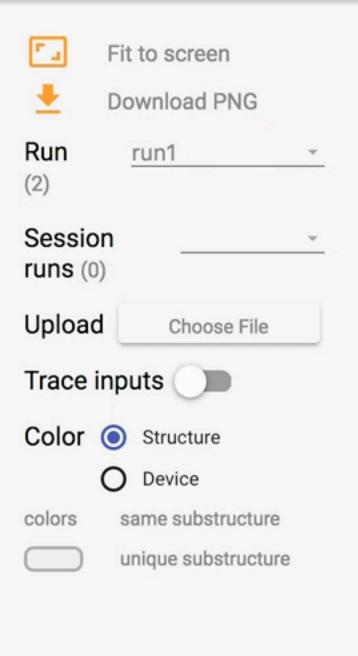


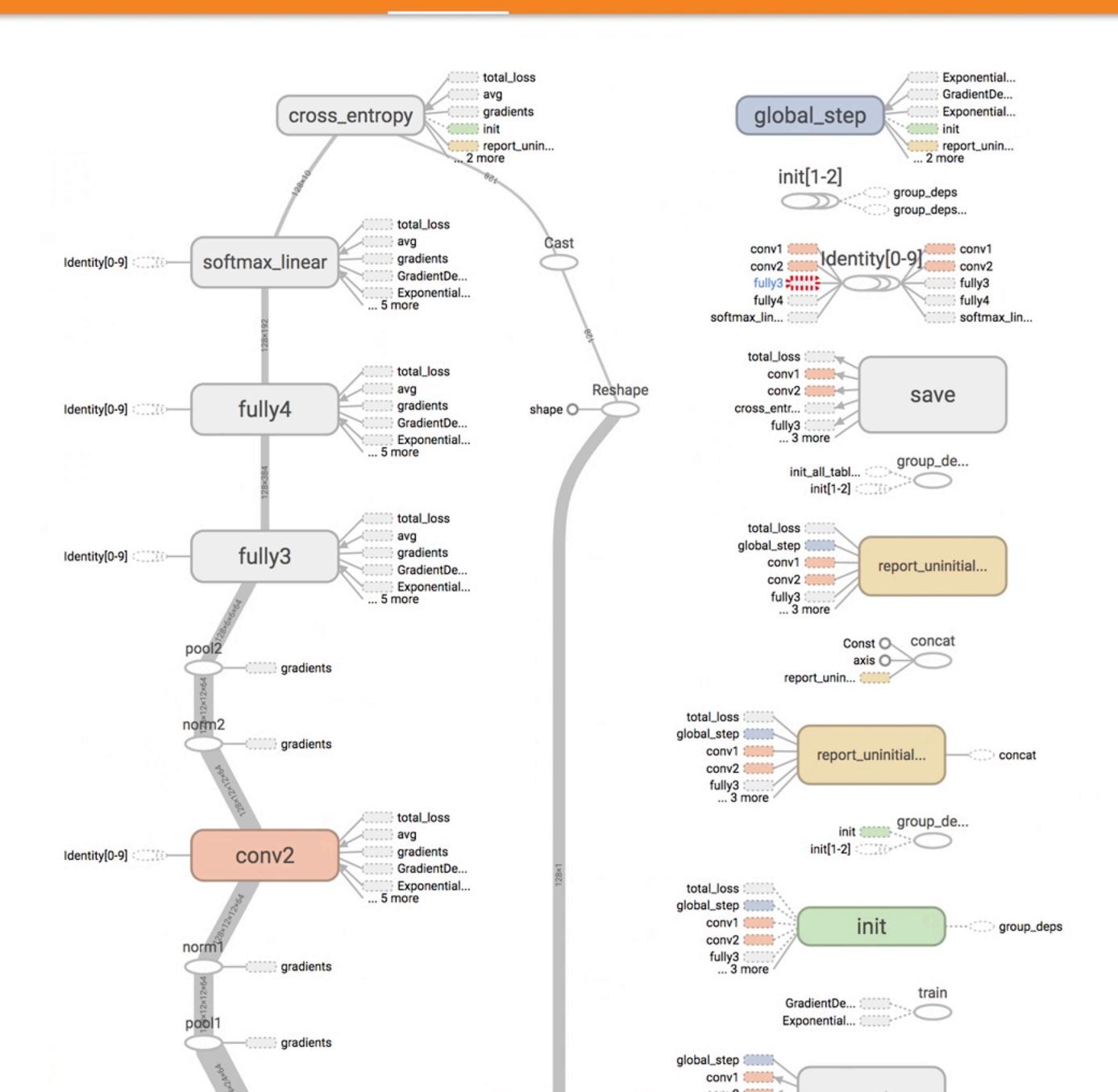




TensorBoard

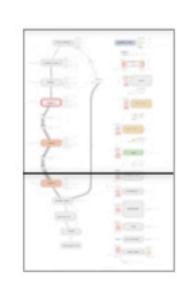
AUDIO











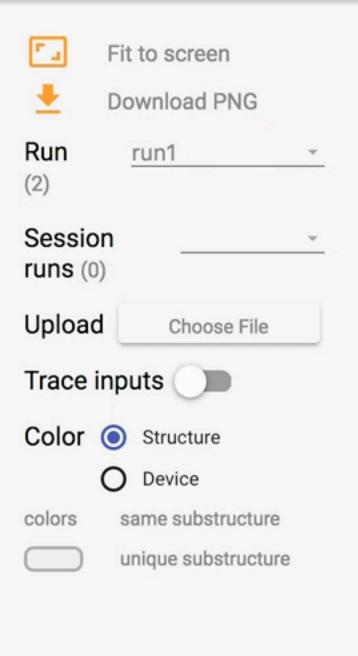
-

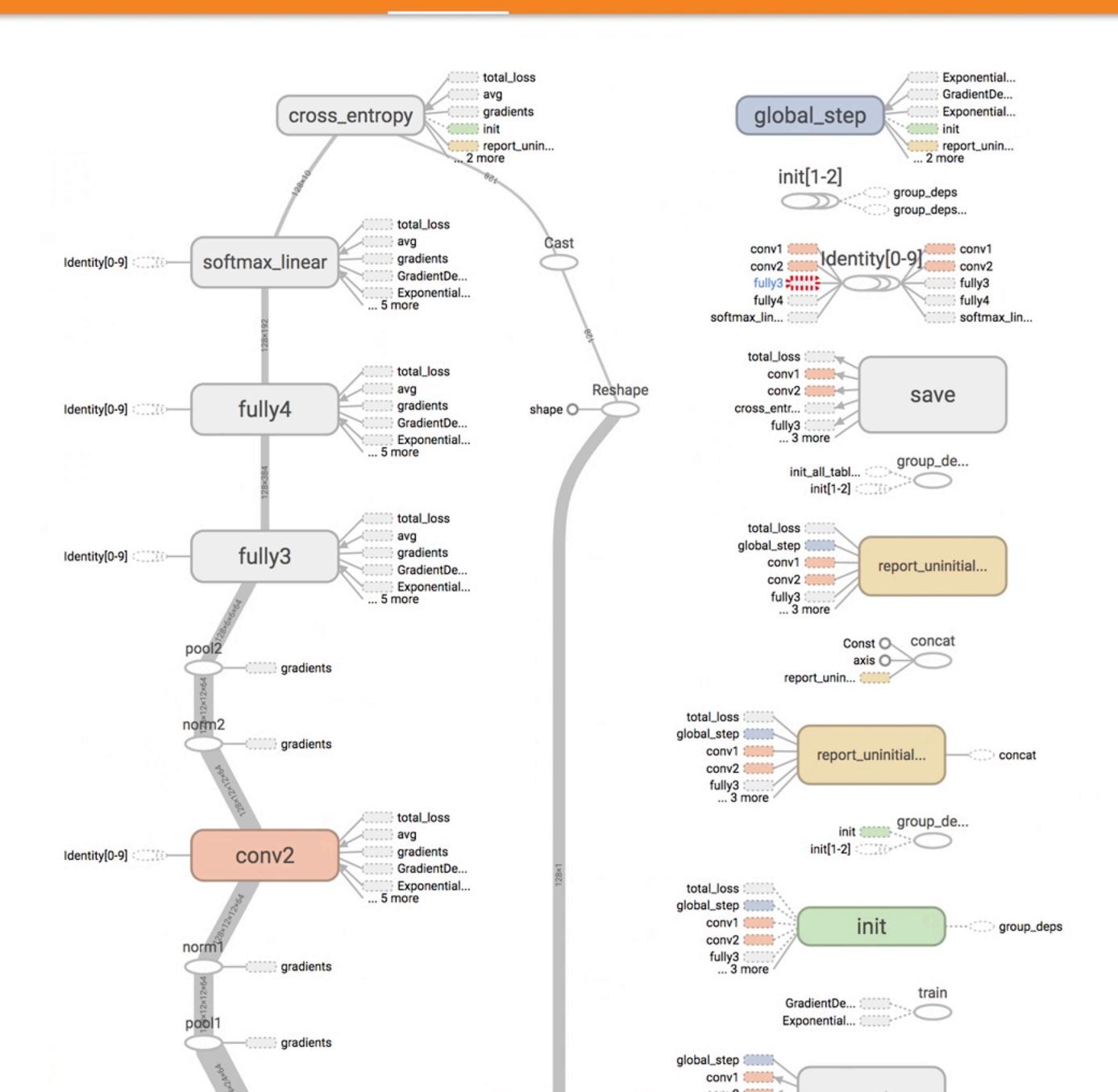
?



TensorBoard

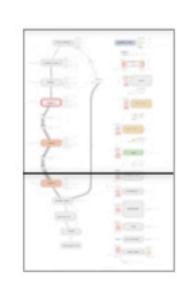
AUDIO







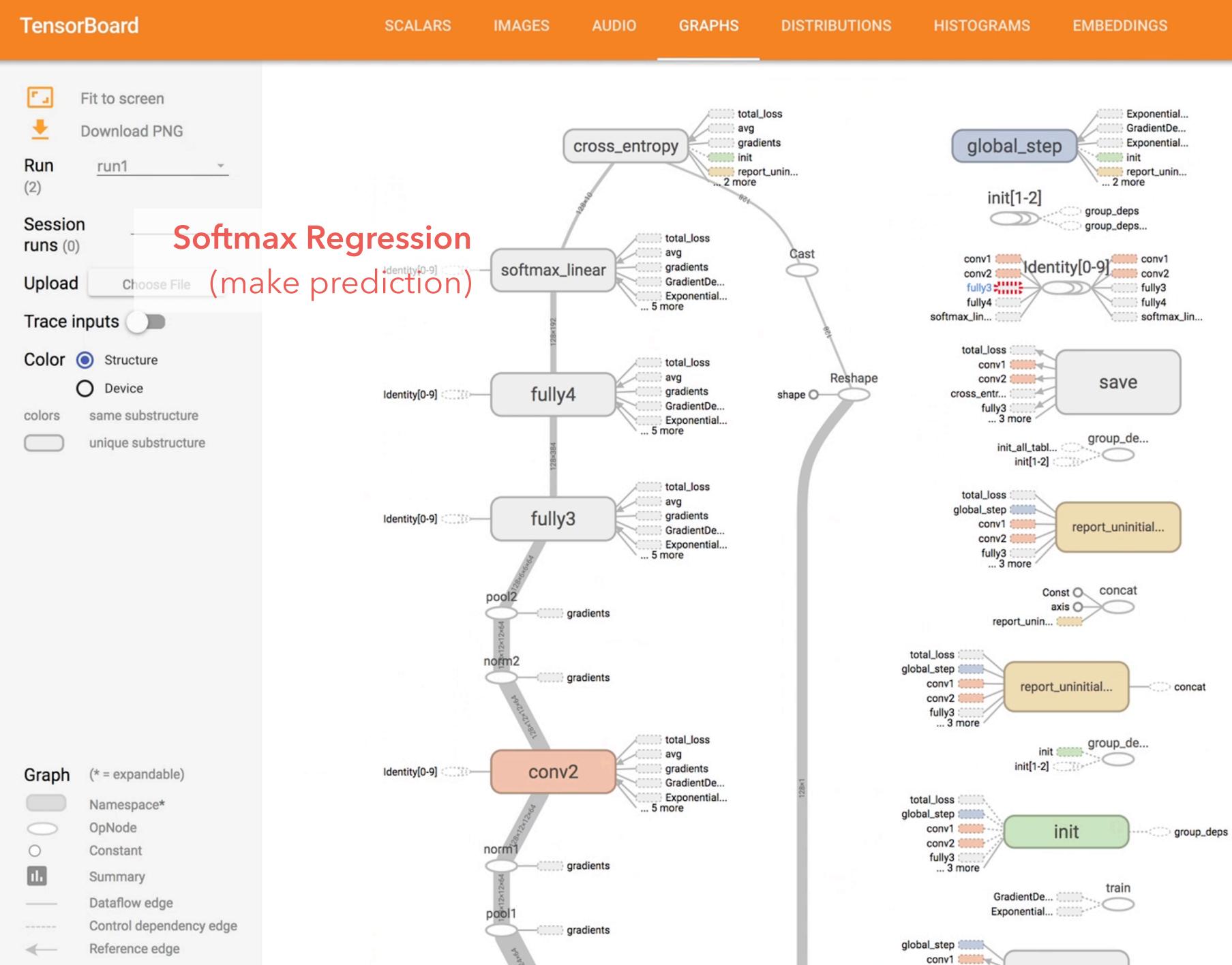




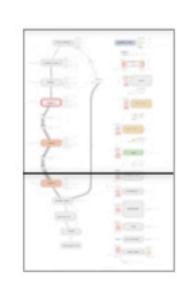
-

?



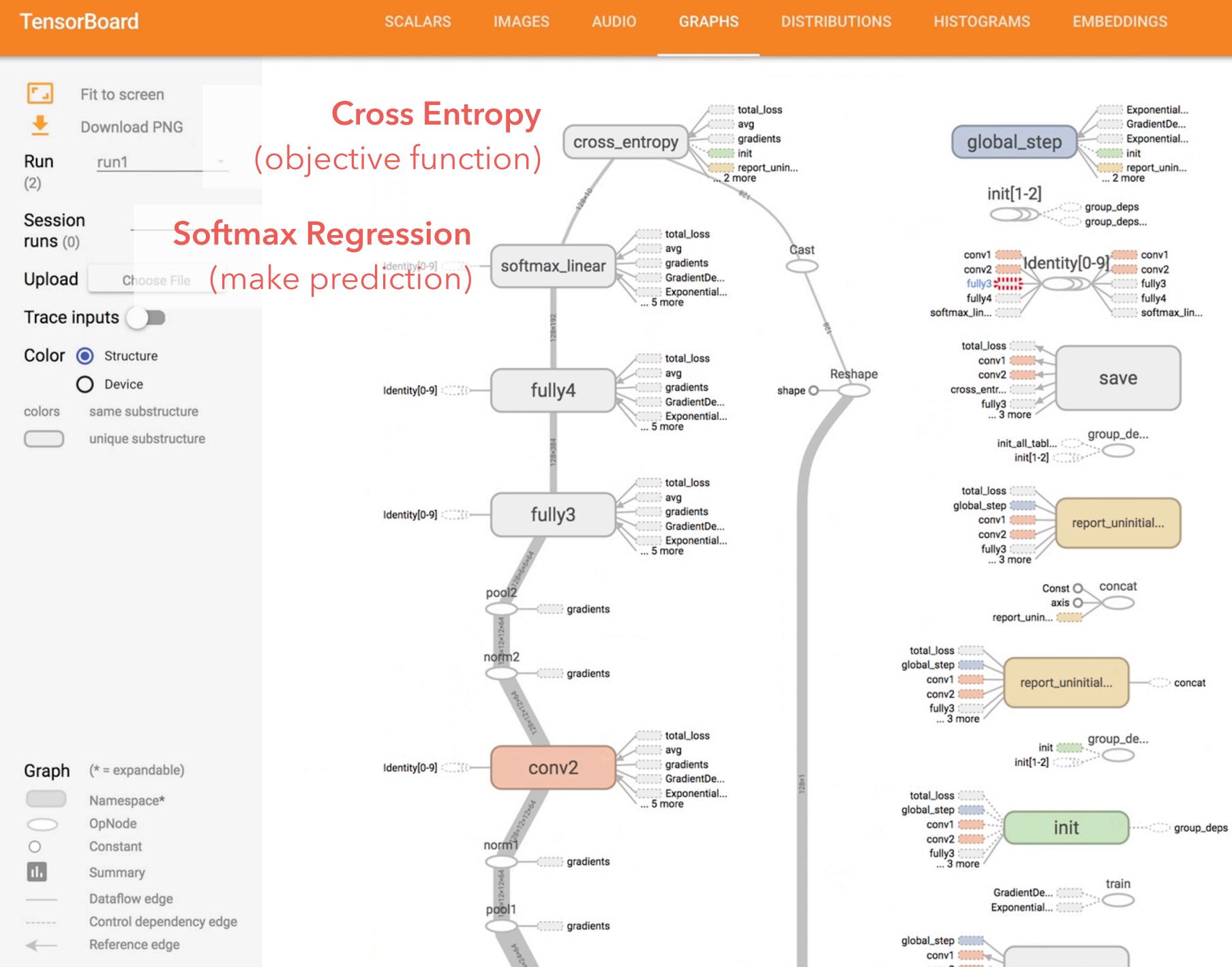




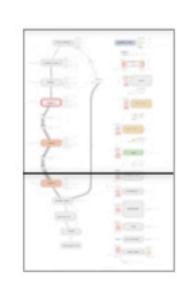


?



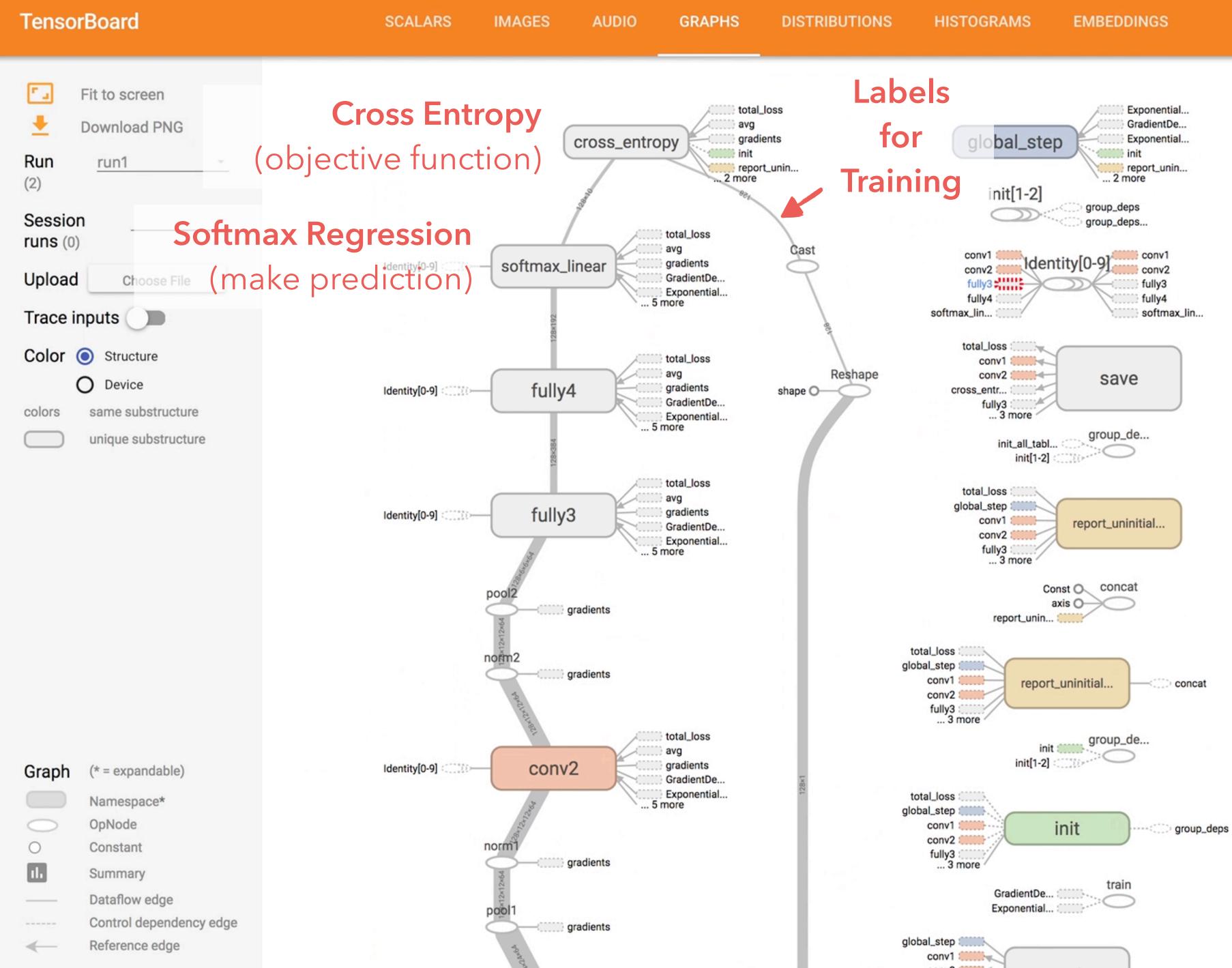


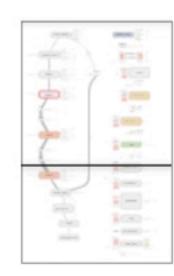




?







?



TensorFlow Graph Visualizer

Visualizing Dataflow Graphs of Deep Learning Models in TensorFlow



TensorFlow Graph Visualizer

Visualizing Dataflow Graphs of Deep Learning Models in TensorFlow

Introduction

Explore a Convolutional Network

Transformation Strategies

Usage Pattern & Feedback



Strategy 1. Extract Less Important Nodes

Related: Dunne & Shneiderman 2013, Van Ham & Wattenberg 2008

Some operations are common, but do not help distinguish different models.



Some operations are common, but do not help distinguish different models.

weights/Assign

zeros





Some operations are common, but do not help distinguish different models.

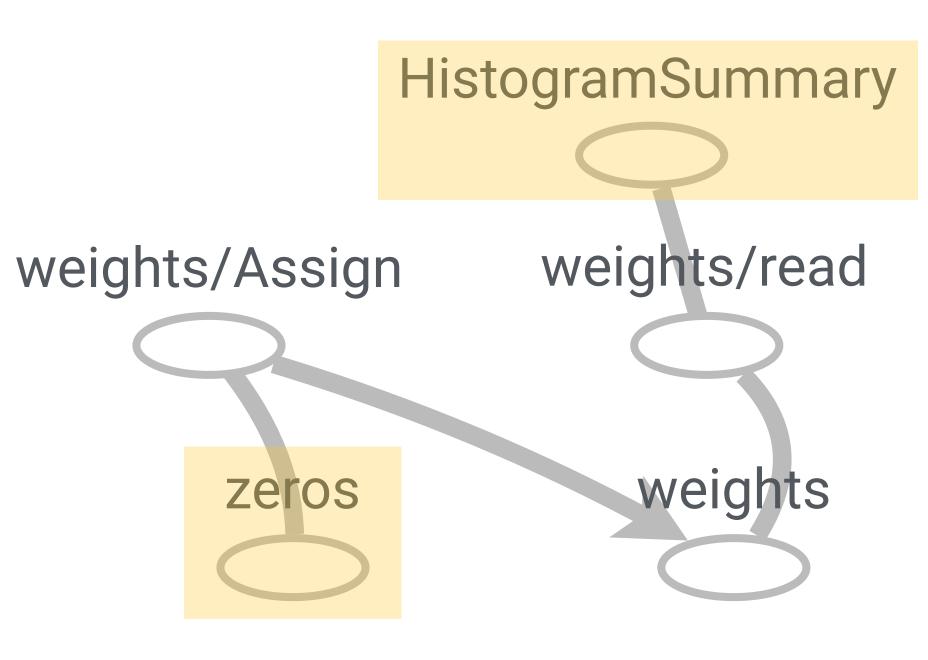
weights/Assign zeros

A constant operation. (considered as a parameter)





Some operations are common, but do not help distinguish different models.

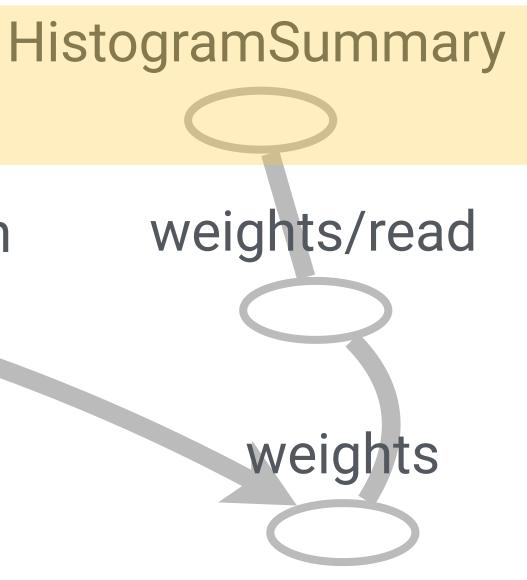


A constant operation. (considered as a parameter)

Summary operations. (for logging data)

Some operations are common, but do not help distinguish different models.

weights/Assign zeros 🔘

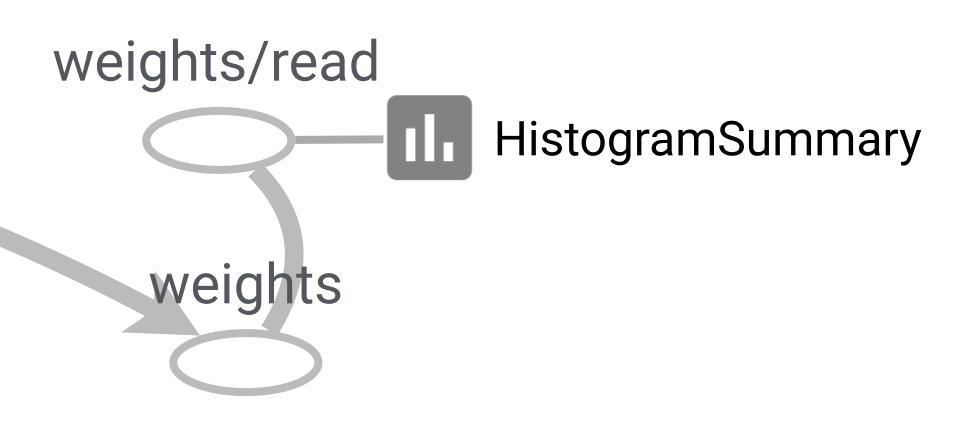


Summary operations. (for logging data)

Some operations are common, but do not help distinguish different models.

weights/Assign zeros 🔘

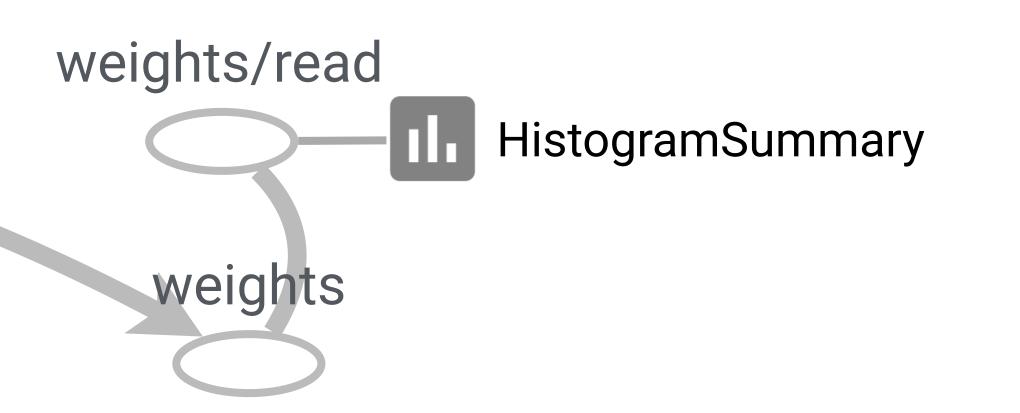




Some operations are common, but do not help distinguish different models.

weights/Assign zeros 🔘





Constants and summaries always connect to only one other operations. Extracting them do not change path between other nodes

Strategy 2. Build a Hierarchical Clustered Graph

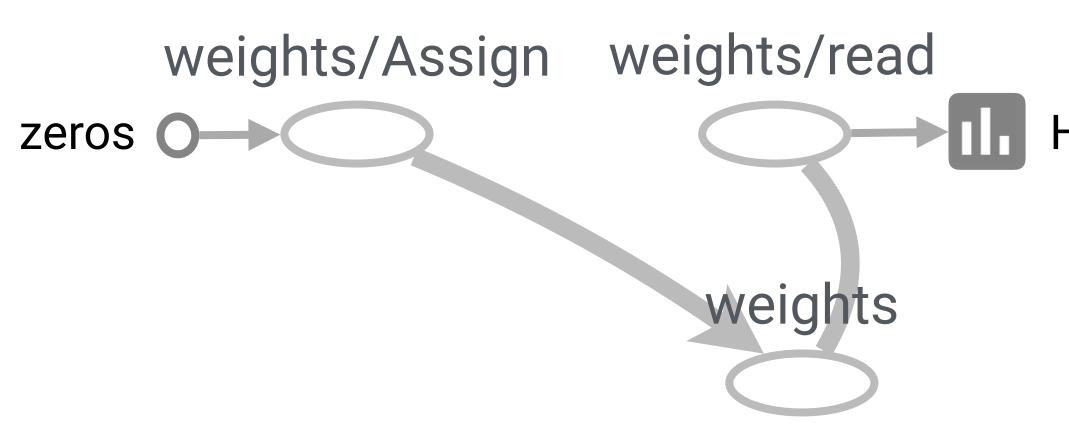
Related: Archambault et al. 2008, Gansner et al. 2005, Balzer et al. 2007

4 🗖

How do we group the nodes?



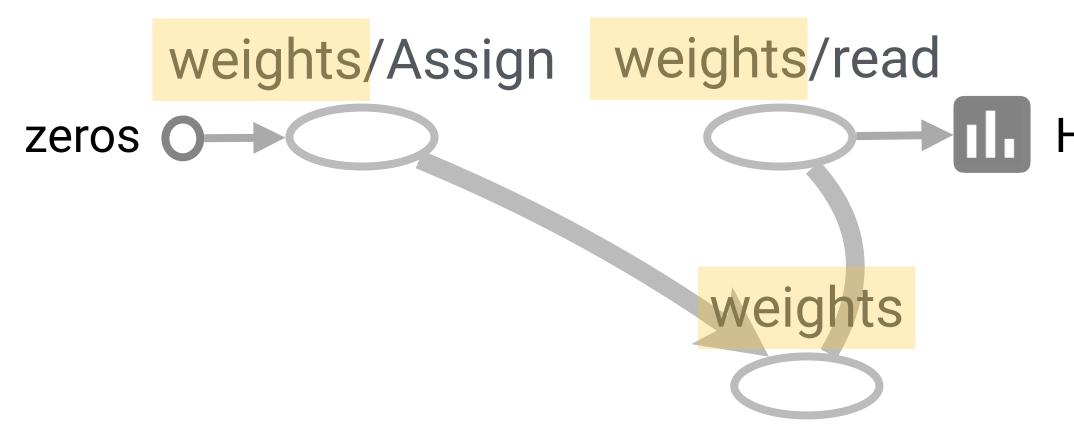
W = tf.Variable(zeros, name='weights')



User can give nodes a *directory-like* namespace (originally for non-visual debugging)

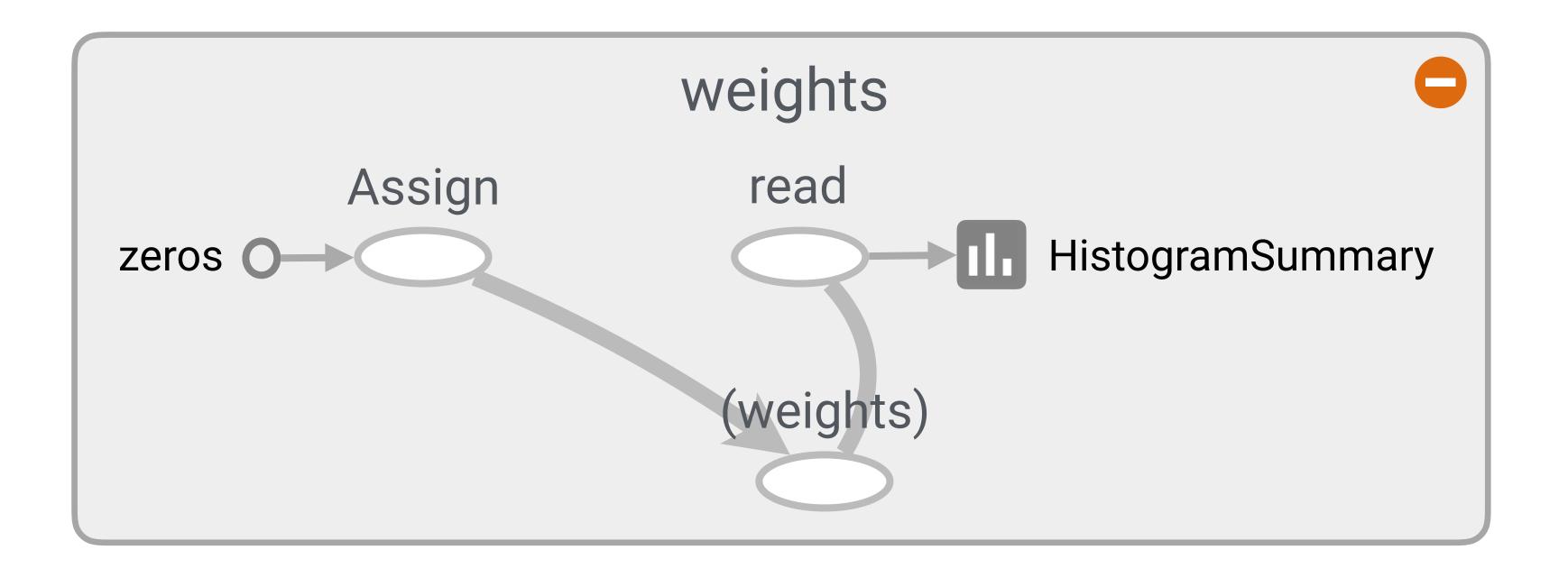
HistogramSummary





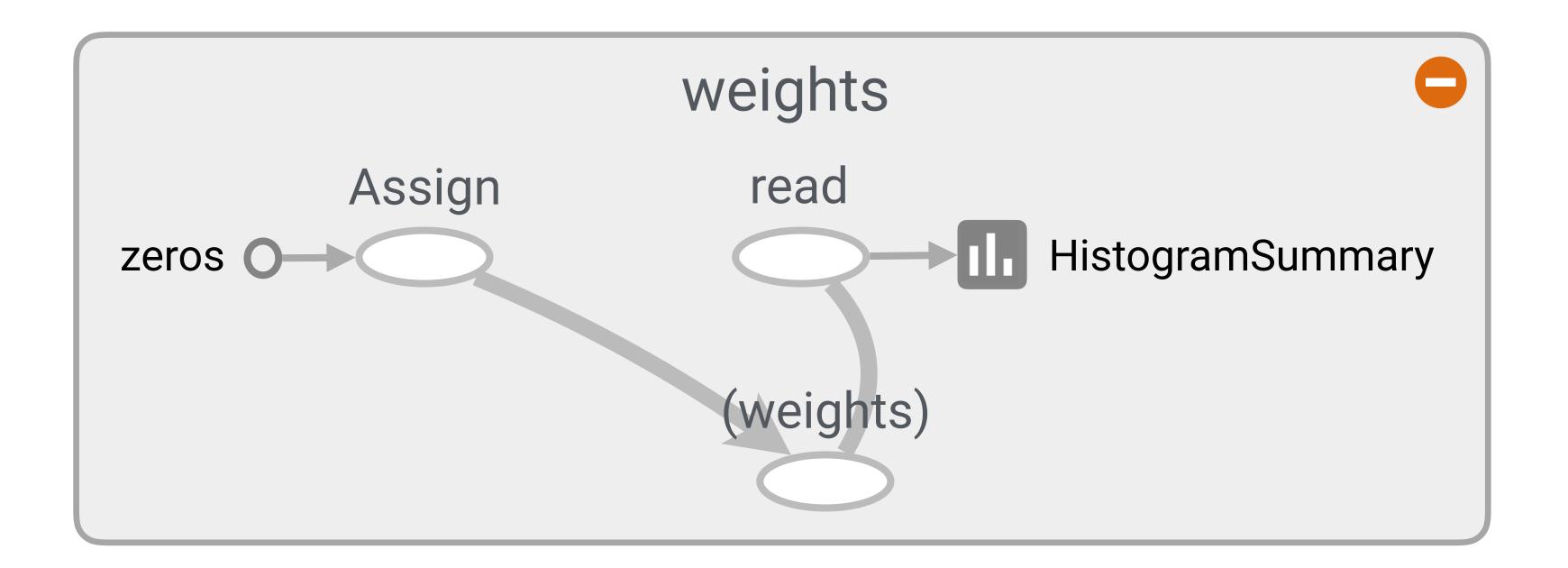
User can give nodes a *directory-like* namespace (originally for non-visual debugging)

HistogramSummary



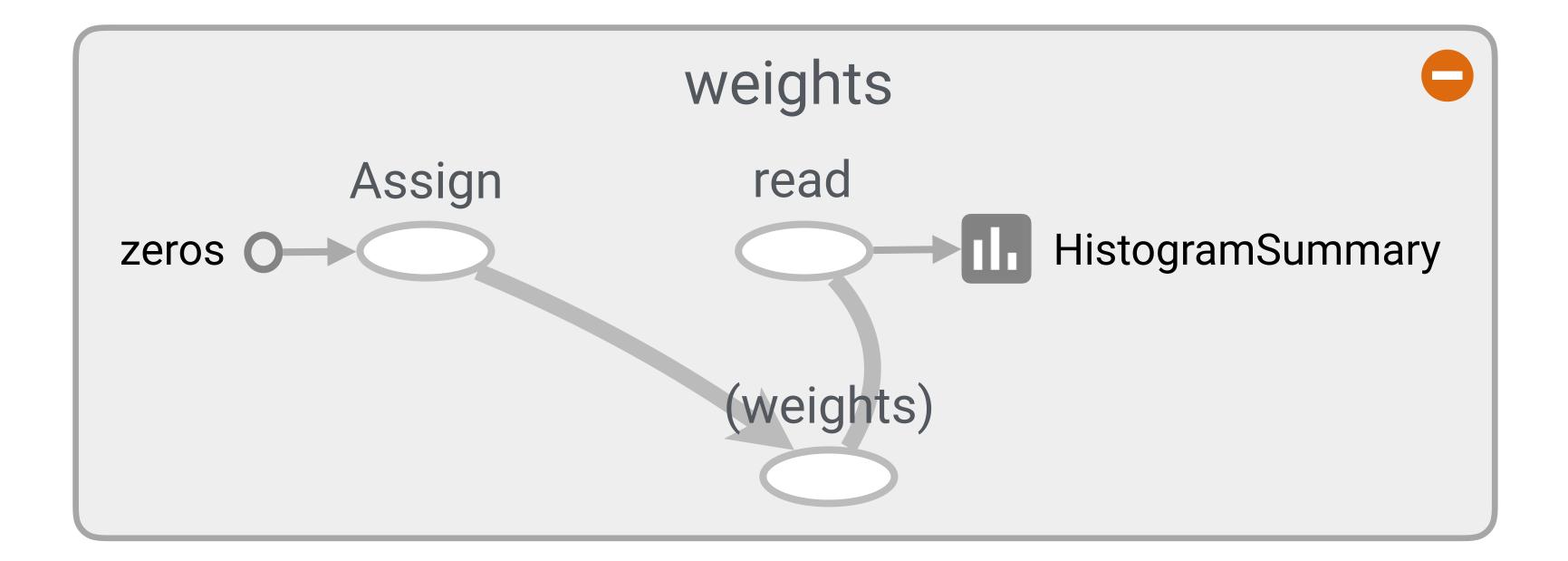


Names are optional but easy to add.

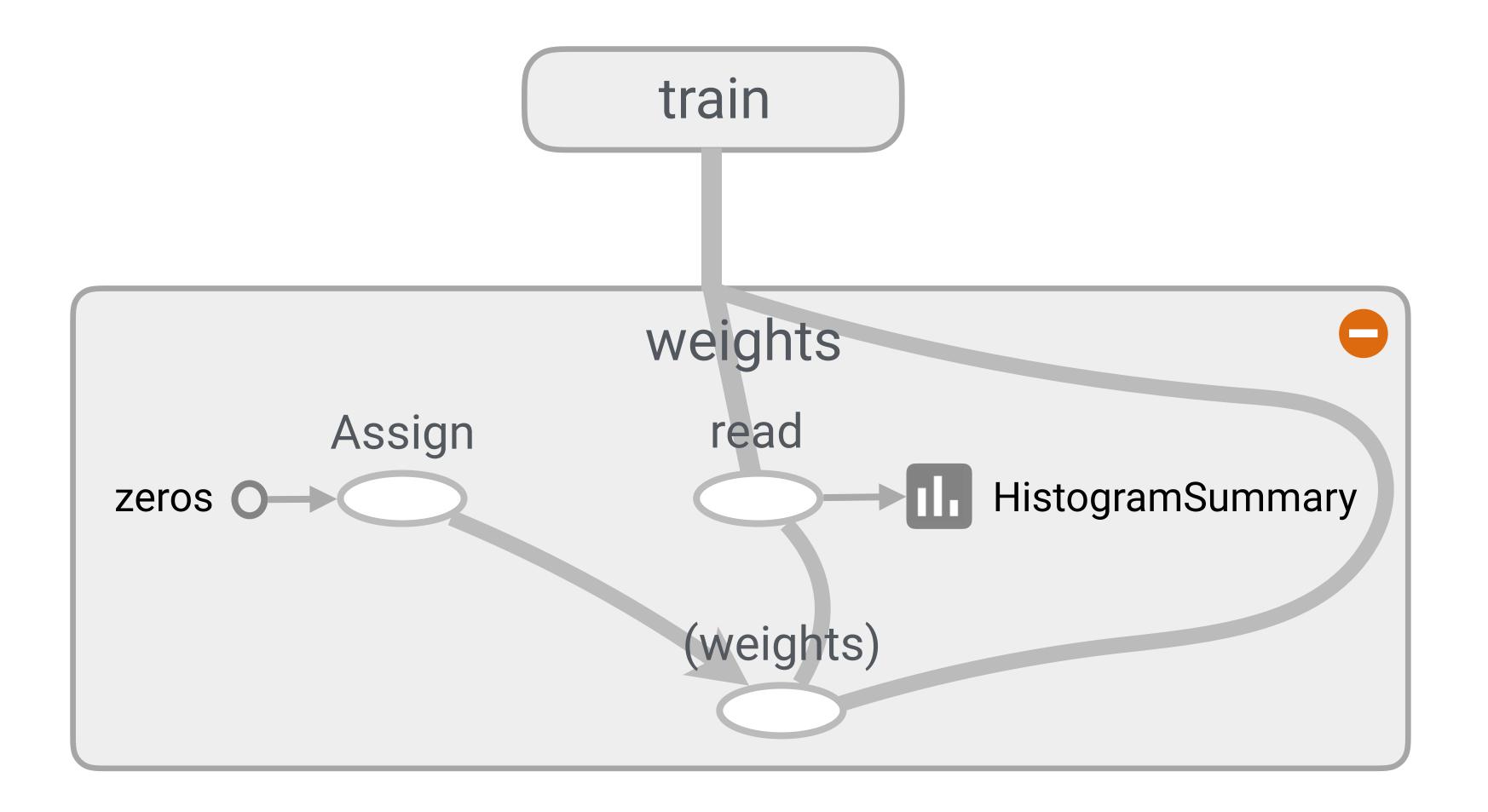


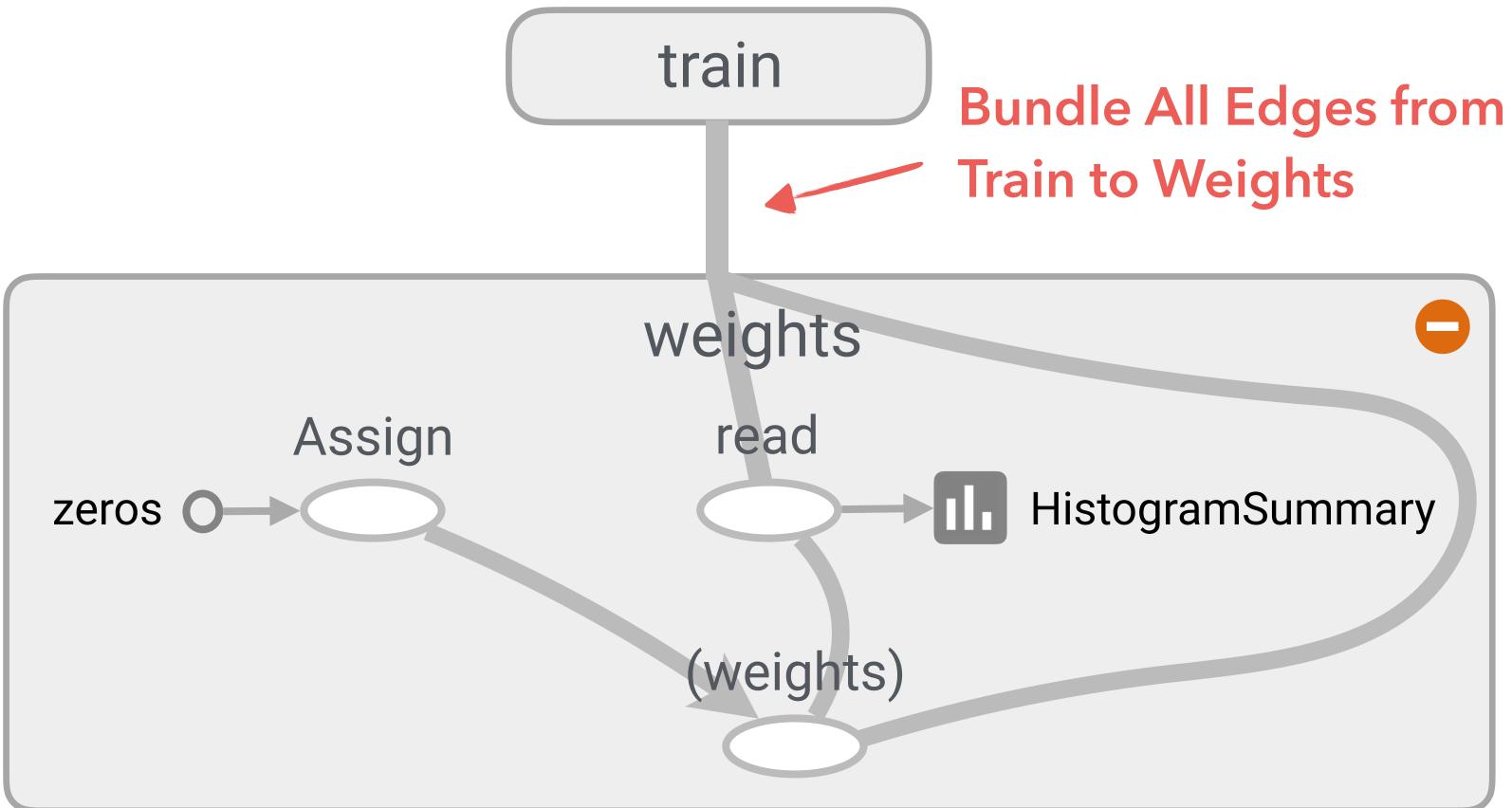


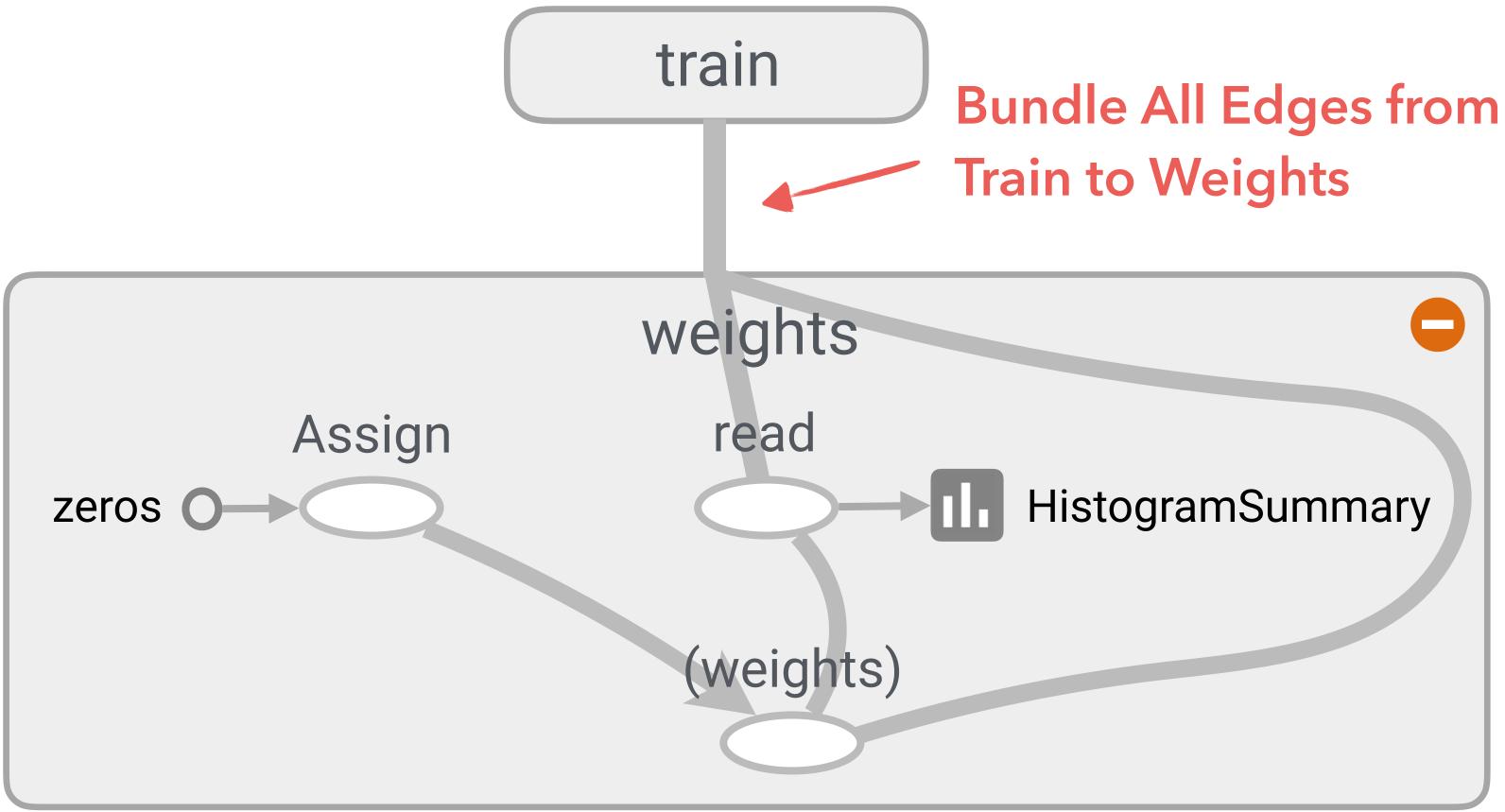
Names are optional but easy to add. Plus, users already used names with non-visual debugging tools.





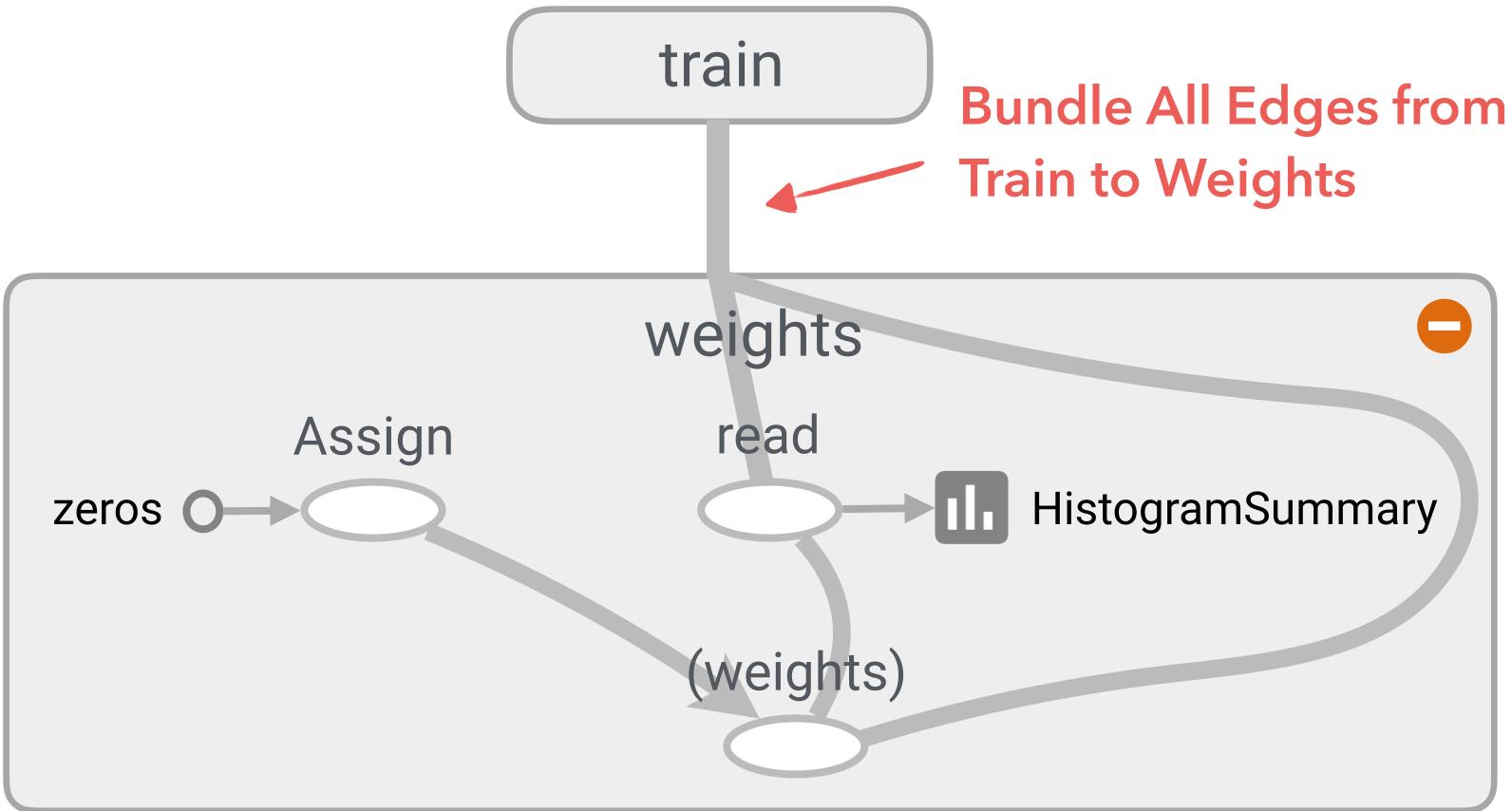






Faster Layout Calculation (recursively calculate layout for each group)

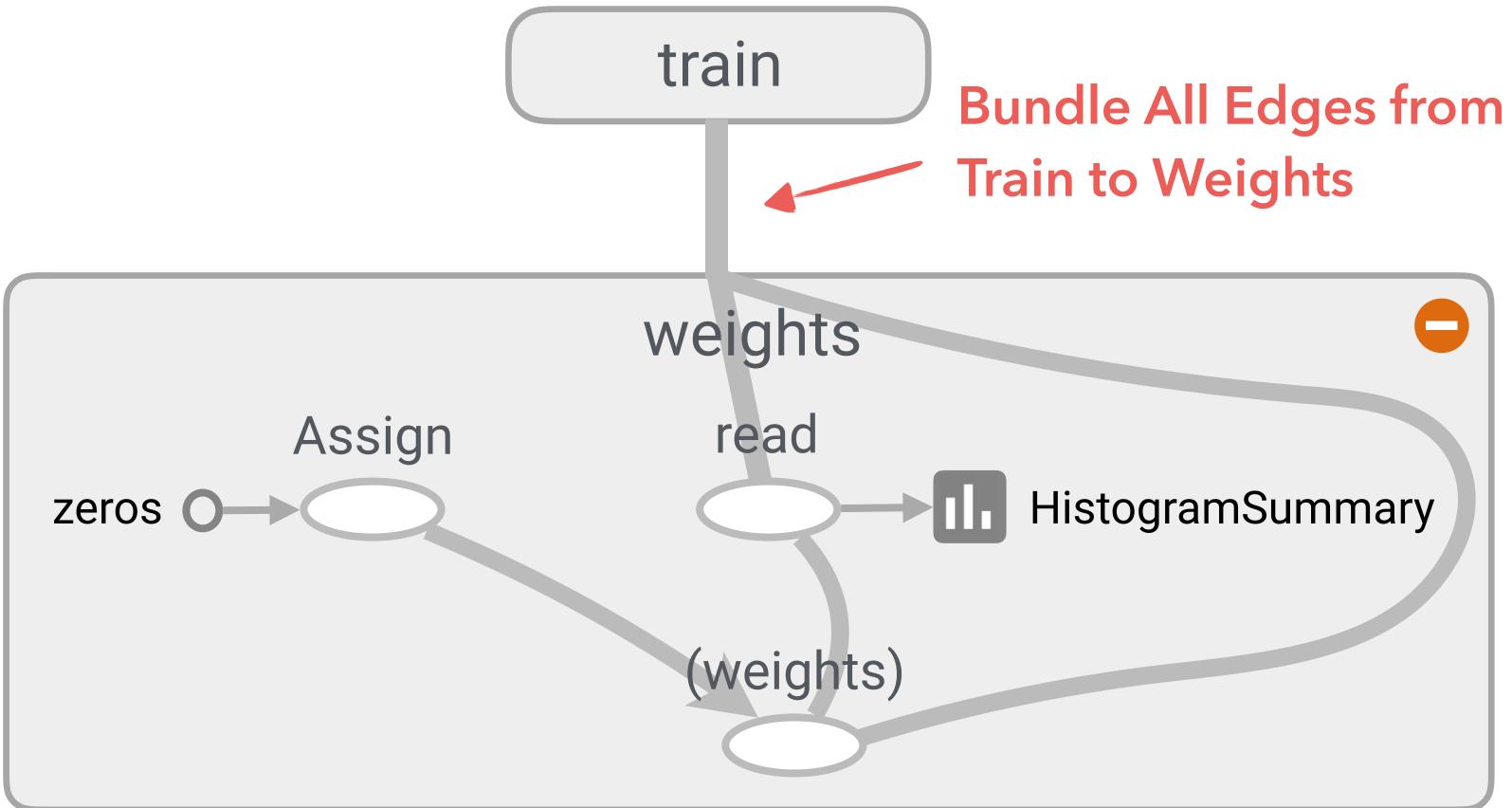




Faster Layout Calculation (recursively calculate layout for each group)

Make layout stable on expansion



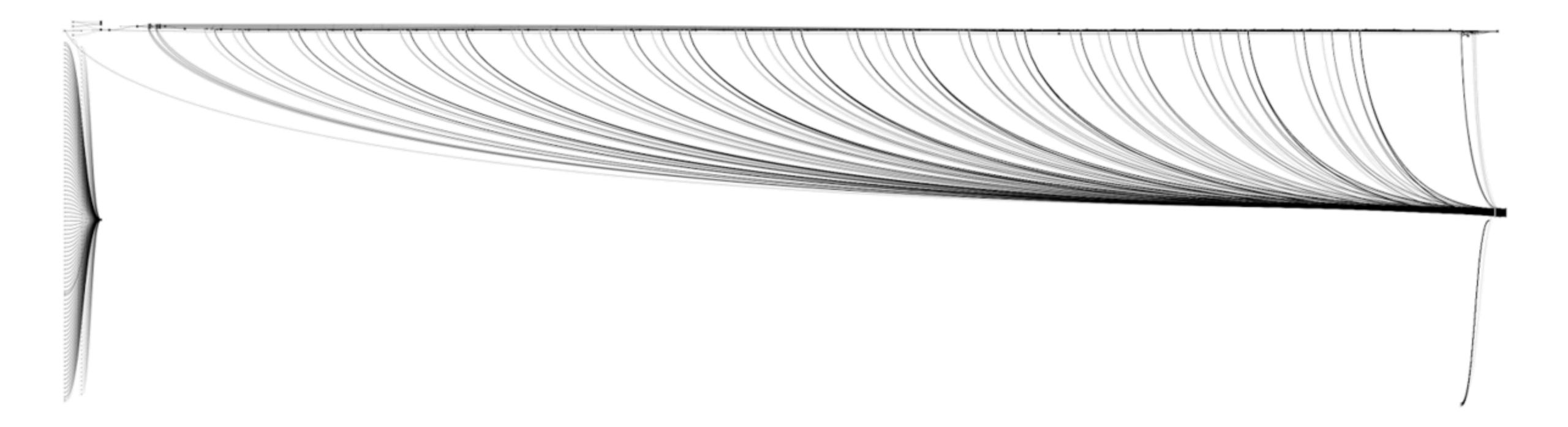


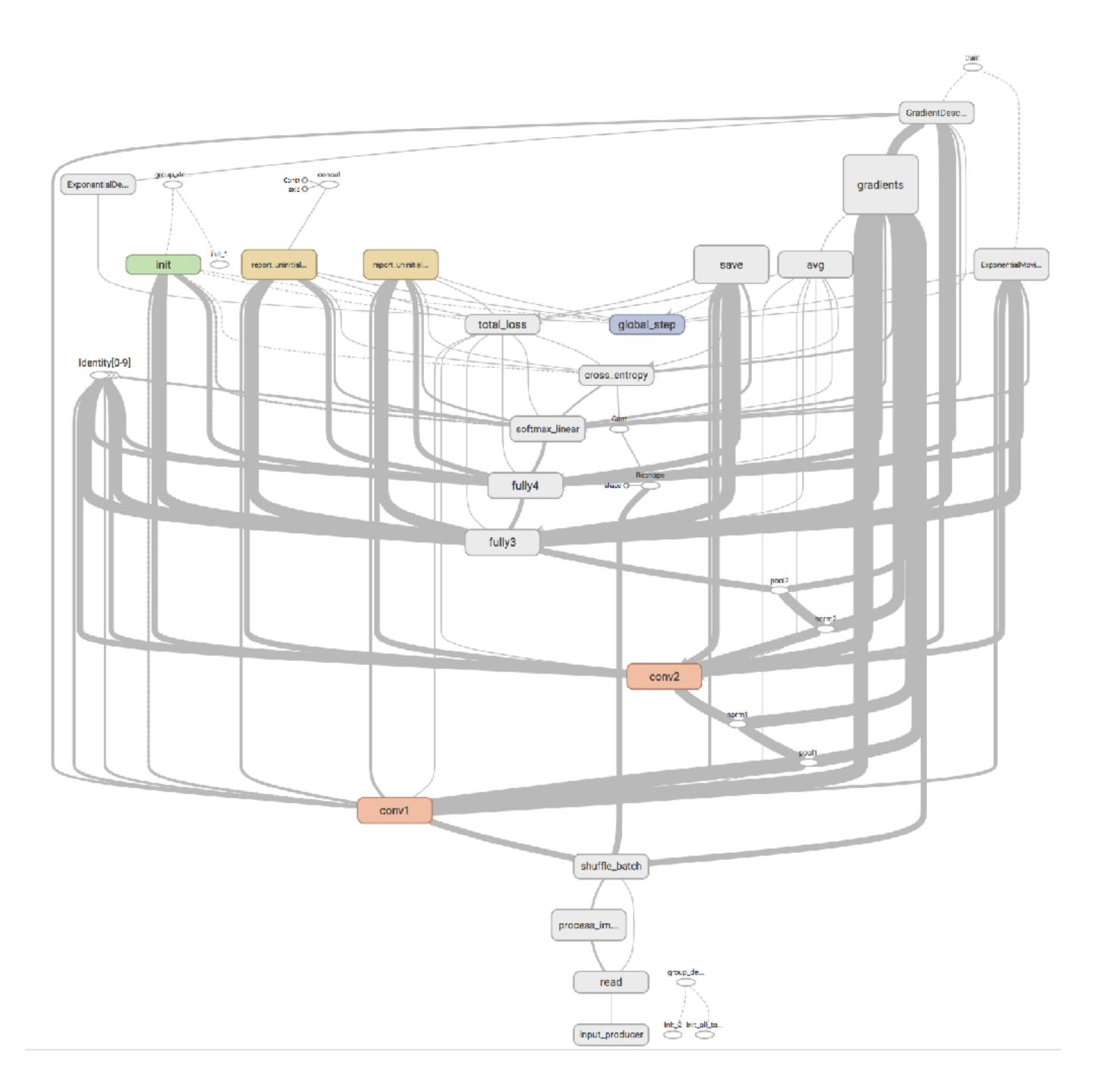
Faster Layout Calculation (recursively calculate layout for each group)

Make layout stable on expansion

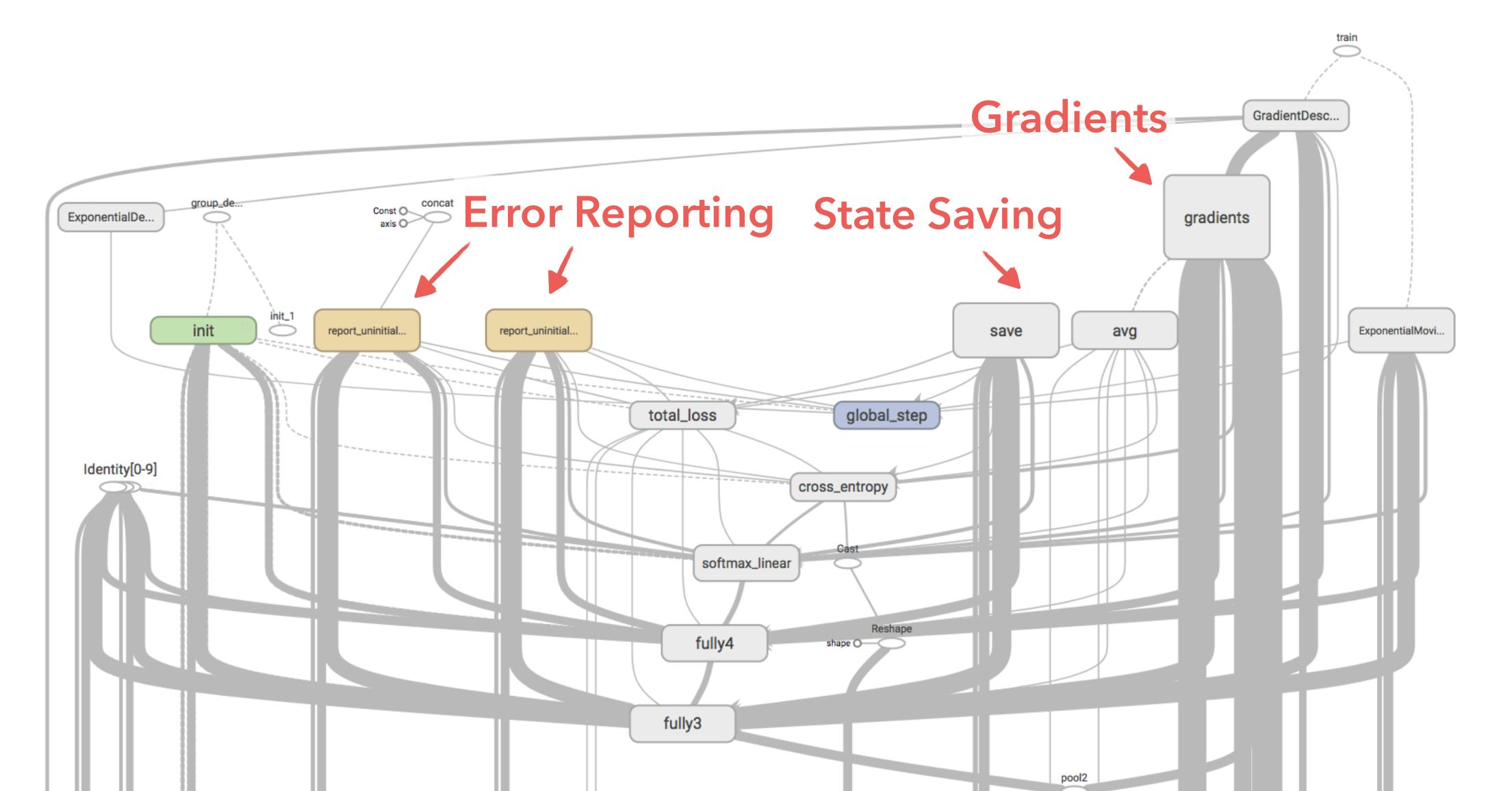
Reduce edge crossing







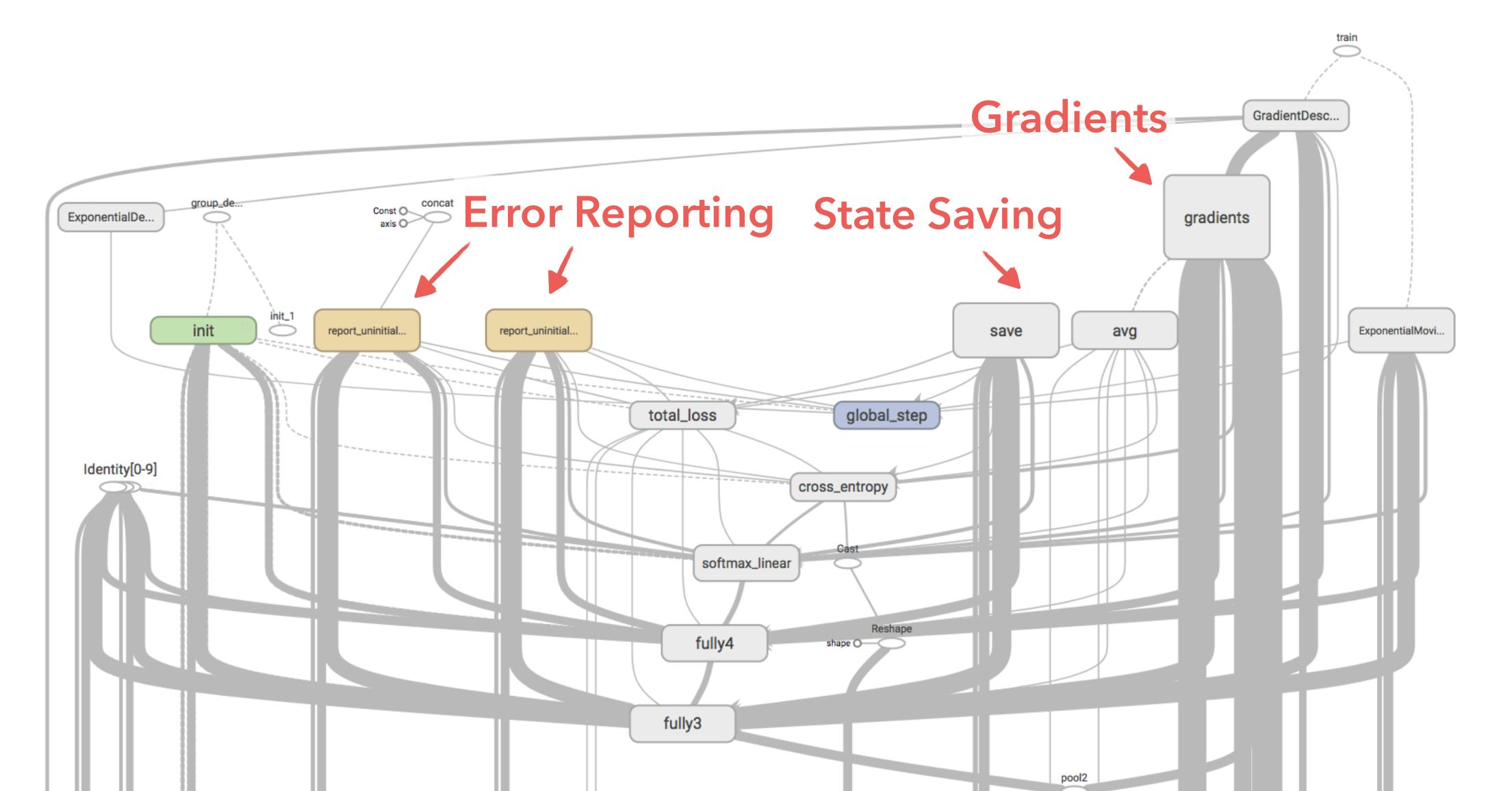
High-degree nodes at the end (and start) connect to all core layers.

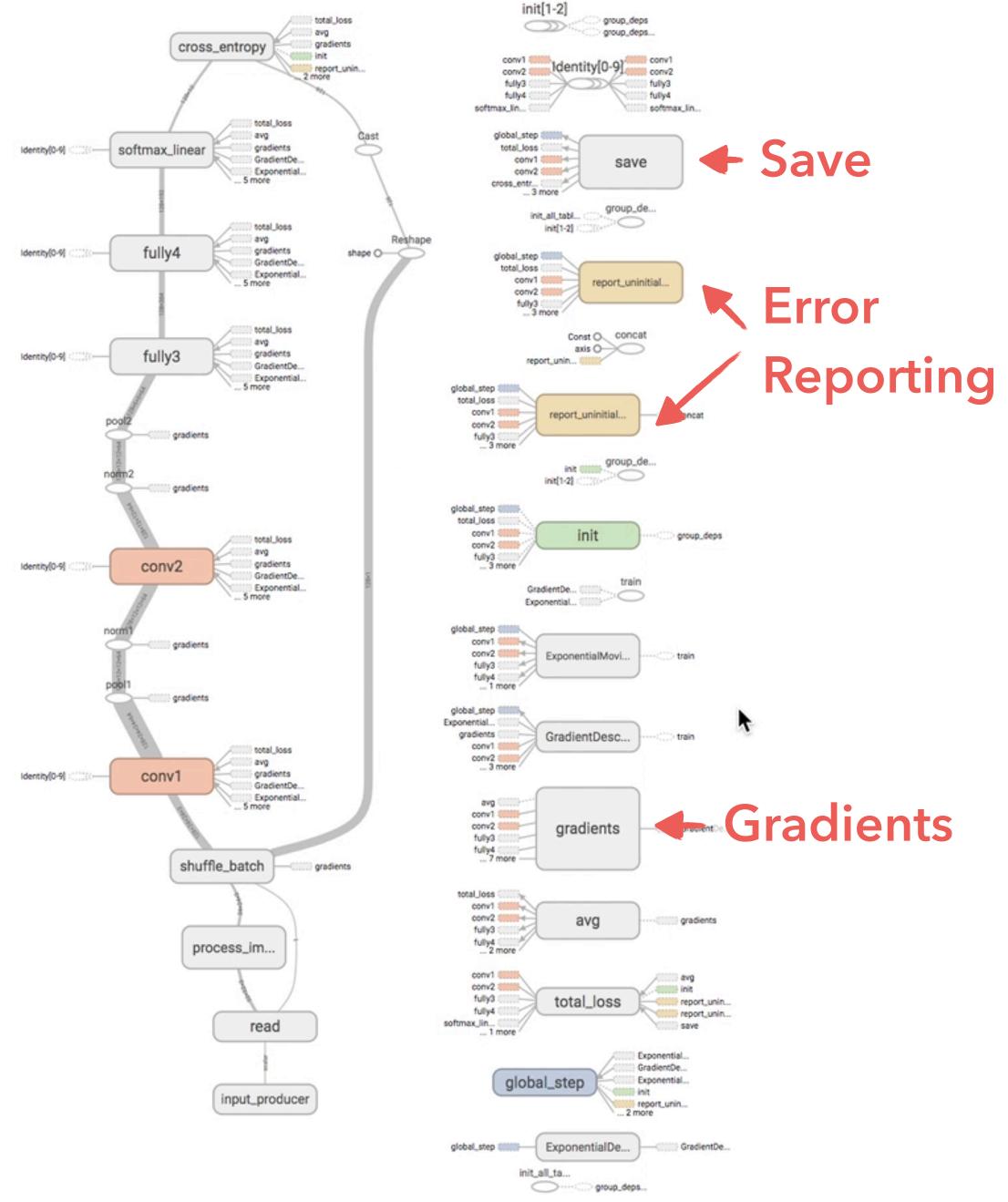


Strategy 1. Extract Less Important Nodes

Strategy 1. Extract Less Important Nodes & Groups

High-degree nodes at the end (and start) connect to all core layers.

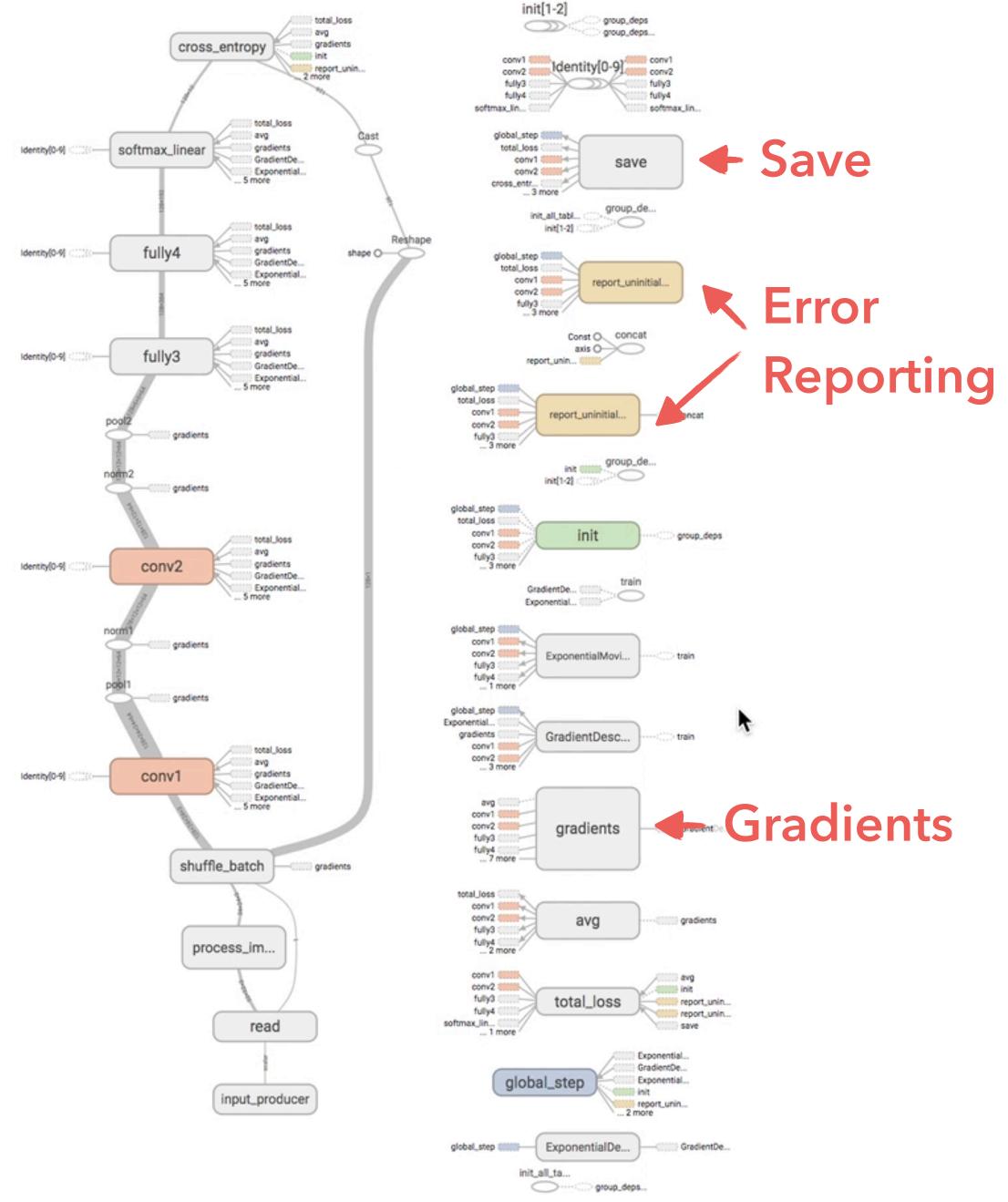




Extract nodes with high in-degree at the end (sinks) & high out-degree at the start (sources) to the right



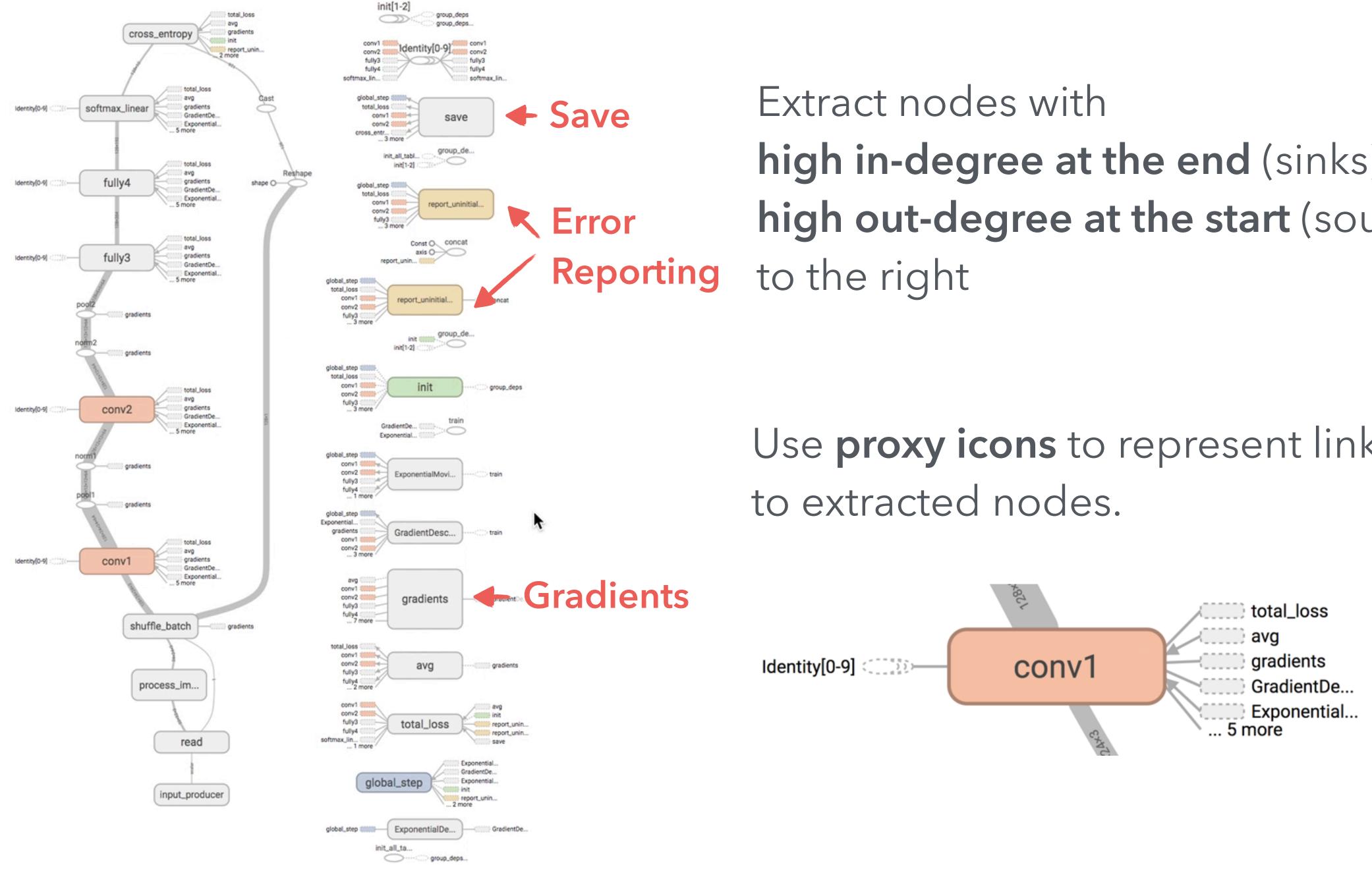




Extract nodes with high in-degree at the end (sinks) & high out-degree at the start (sources) to the right





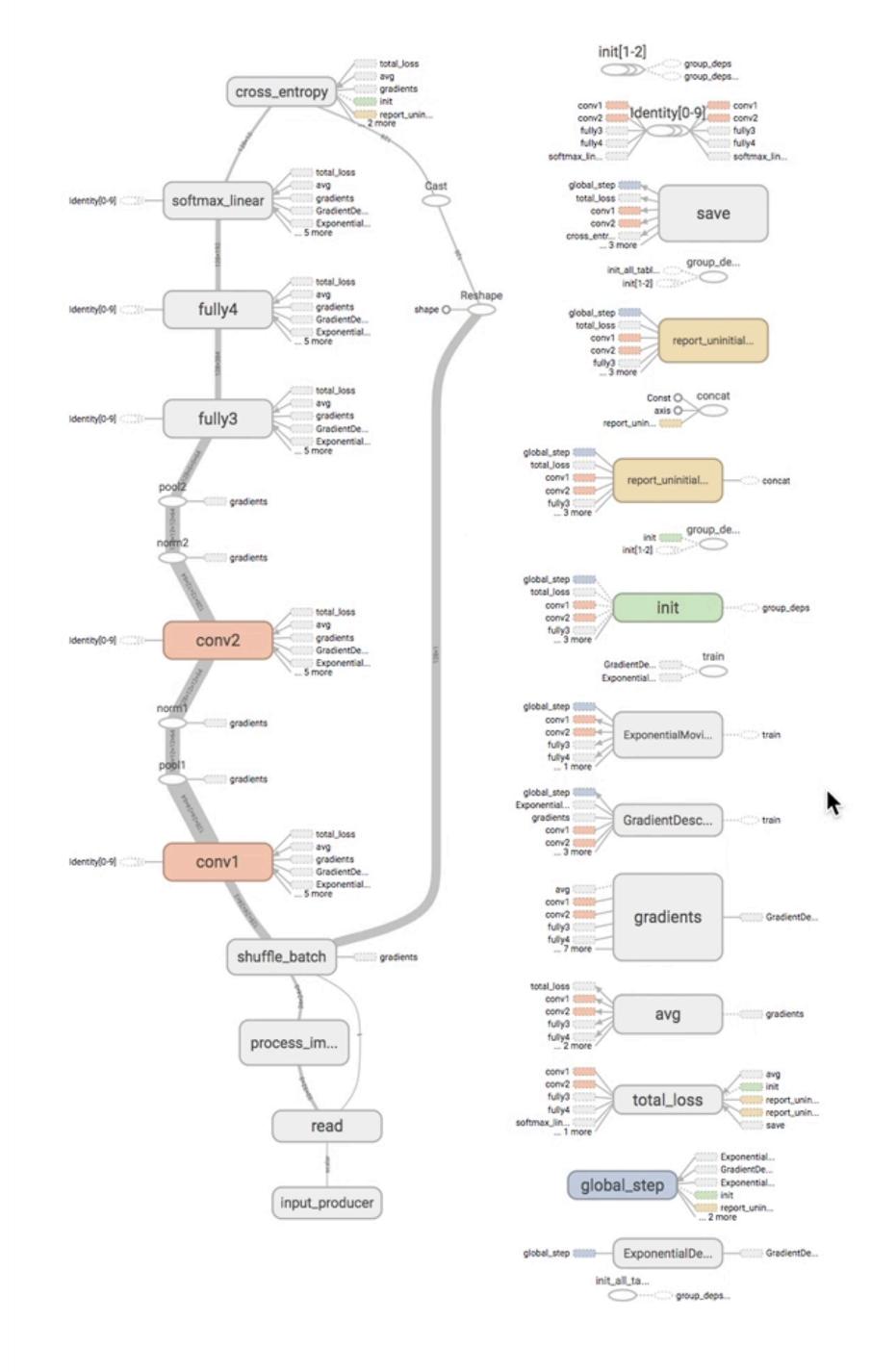


high in-degree at the end (sinks) & high out-degree at the start (sources)

Use **proxy icons** to represent links

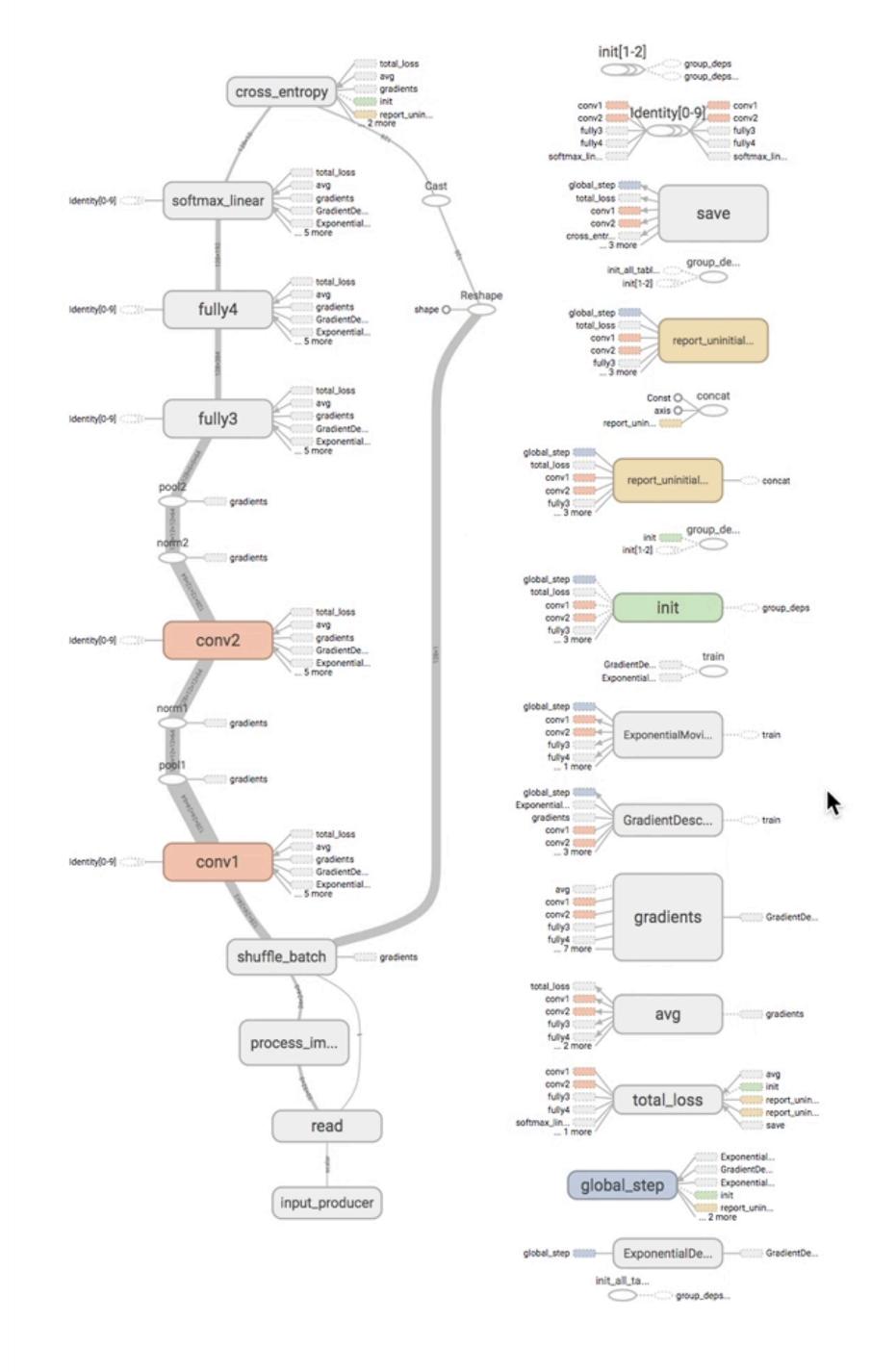






Users can **add nodes back to the main graph** or **extract more nodes**





Users can **add nodes back to the main graph** or **extract more nodes**



Visualizing Dataflow Graphs of Deep Learning Models in TensorFlow



Visualizing Dataflow Graphs of Deep Learning Models in TensorFlow

Introduction

Explore a Convolutional Network

Transformation Strategies

Usage Pattern & Feedback



Visualizing Dataflow Graphs of Deep Learning Models in TensorFlow



Visualizing Dataflow Graphs of Deep Learning Models in TensorFlow

Introduction

Explore a Convolutional Network

Transformation Strategies

Usage Pattern & Feedback



Feedback Sources

- 1) Structured questionnaire for internal users at Google
- 2) Mailing list conversations
- 3) Public feedback from online articles



Usage Pattern: Inspecting New Models

Usage Pattern: Inspecting New Models

"Understand what my code actually produced. We had layers of functions and configs... it's good to verify that we got what we intended"

Usage Pattern: Inspecting New Models

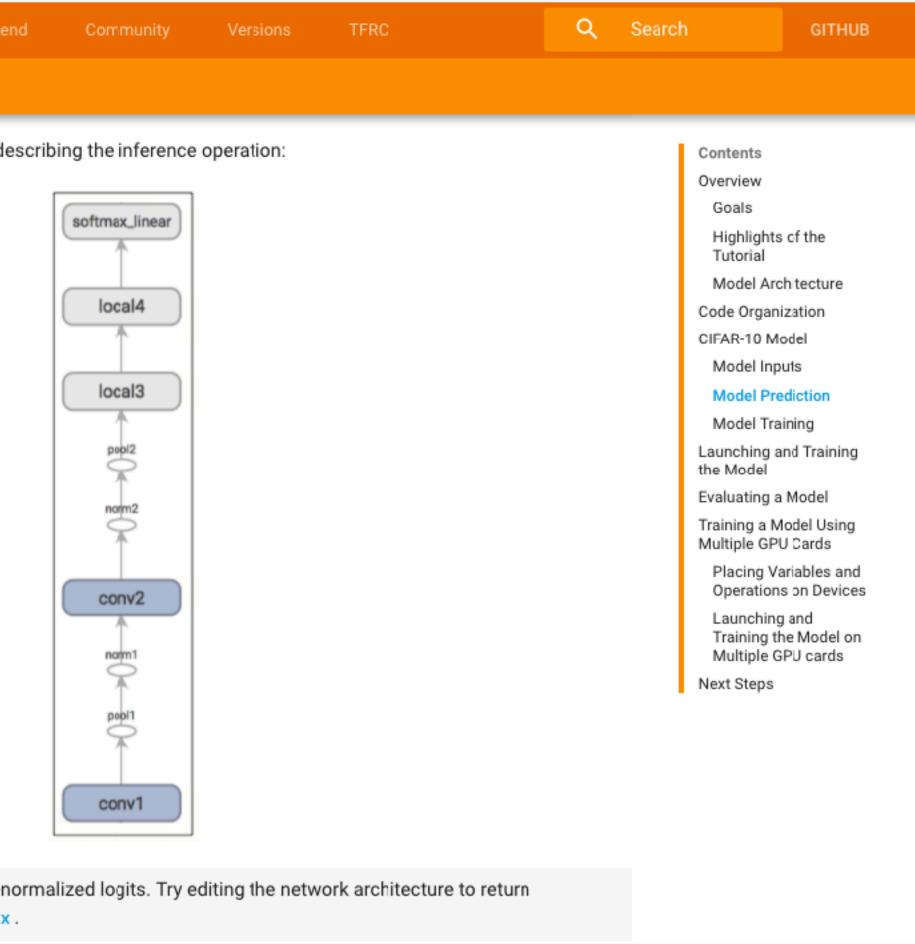
"Understand what my code actually produced. We had layers of functions and configs... it's good to verify that we got what we intended"

"Find the name of a tensor so that I could do further exploration (like seeing the evolution of a particular input)"

Usage Pattern: Using Screenshot to Explain Models

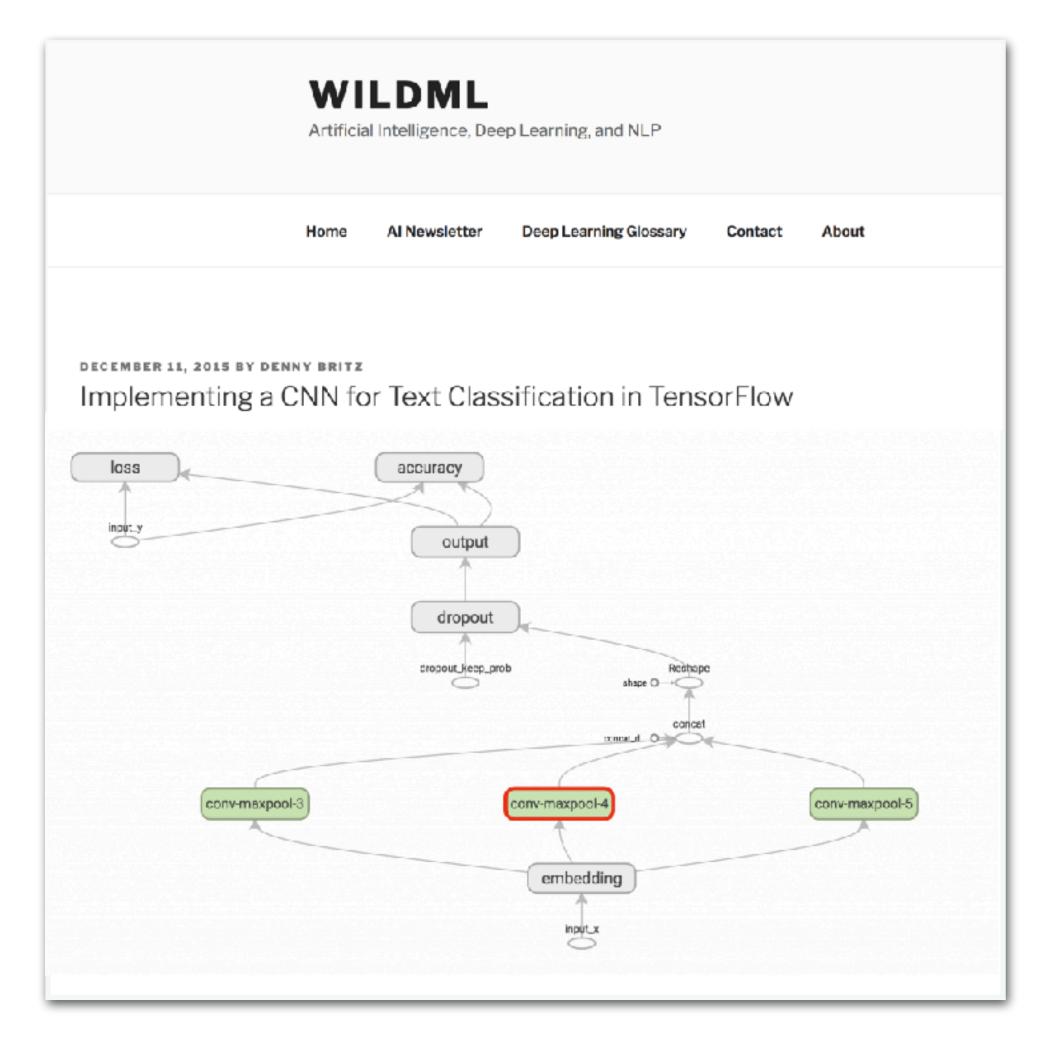
TensorFlov	N M	Install	Develop	API r1.3	Deploy	Extend
GET STARTED	PROGRAMMER	'S GUIDE	TUTORIALS	S PER	FORMANCE	
			Here is a graph	generated	from Tensor	Board des
Tutorials						
Using GPUs						
Image Recognition	I					
How to Retrain Inco for New Categories						
A Guide to TF Laye Convolutional Neur						
Convolutional Neu	ral Networks					
Vector Representat	tions of Words					
Recurrent Neural N	letworks					
Sequence-to-Seque	ence Models					
Large-scale Linear TensorFlow	Models with					
TensorFlow Linear	Model Tutorial					
TensorFlow Wide & Tutorial	Deep Learning					
Improving Linear M Kernel Methods	Iodels Using Explic	it				
Mandelbrot Set						
Partial Differential	Equations					
TensorFlow Version	ns					
			EXERCISE: The normalized pre	-		

TensorFlow's Official Tutorials





Usage Pattern: Using Screenshot to Explain Models



3rd Party Articles

Questions	Developer Jobs	Tags	Users	Search	
	2_hidden	ted cor	rectly to the	rest of my grap	h
param	uvals vals p_prob_later_val	or or	keep_prob		keep_prob_input_val Sub x O→ Sub ScalarSu_
global_s (ExponentialDecay	9	obal.s	xponentialDecay	global_s ()))) -> ExponentialDecay
Is this the e	-	ior? Is t	nere a way	-	e TensorBoard visualization of my sponding placeholders?
computed v		ectly. Fo	or example,		by show them? And why other nomentum values, which are

StackOverFlow Questions



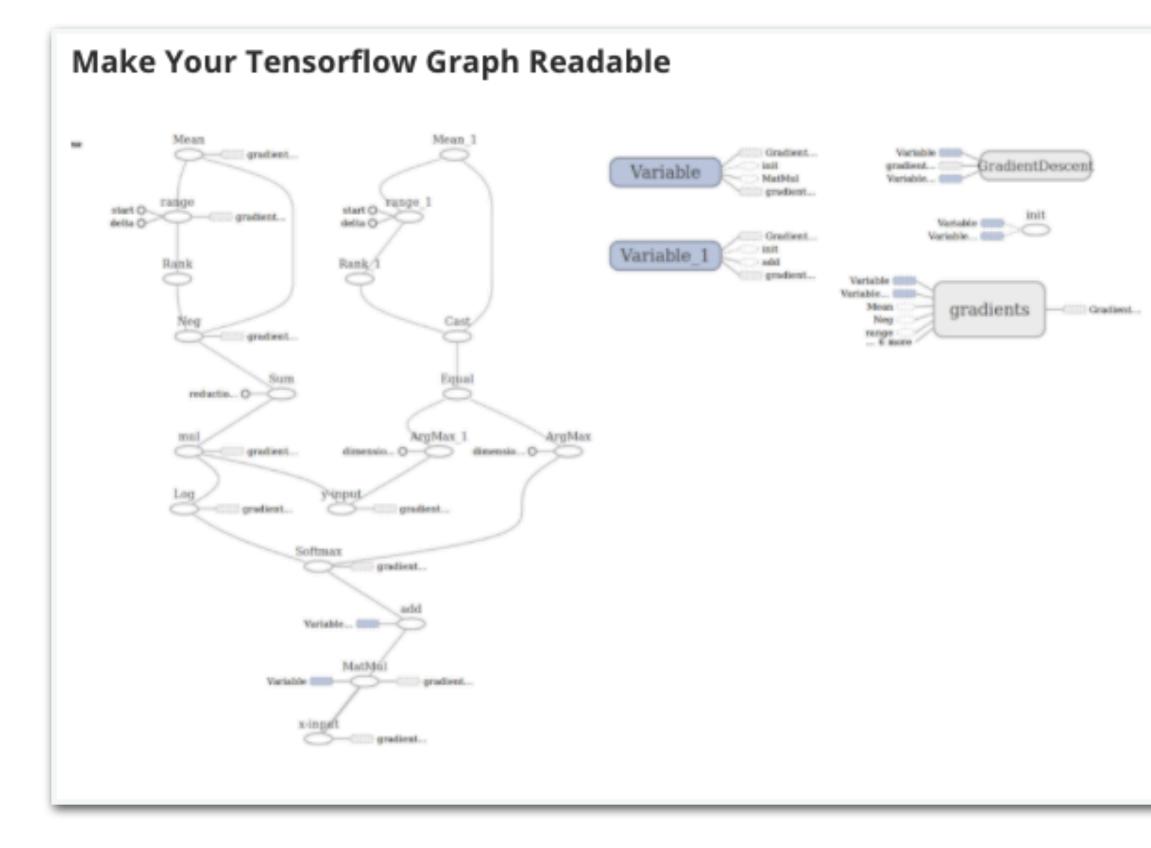
Usage Pattern: Rename Nodes to Improve Visualization

Many users iteratively rename until the visualization match their mental model, especially when sharing with others.



Usage Pattern: Rename Nodes to Improve Visualization

Many users iteratively rename until the visualization match their mental model, especially when sharing with others.



File Machine Yiew Input Devices Help TensorBoard - Mozilla Firefox			ti 🗈 📼
(a) a localhost: 6006	•	C Search	公白 🛡 🕸
TensorBoard	EVENTE BALLEPS A	LEND GRAIN-IS DISTRIBUTIONS HISTOGRAMS	
Pitto screen.			transformation Subgraph: 2 node
			Attributes (0)
Session -	A	в	Inputs (0)
Upload Disself in Trace inputs (39)	y O->	y O->	Outputs (0)
Color @ Structure			Remove from man
allas simesubstruture	Add	bbA	
	× O	×0	
Graph (* expandiable)			
Doomenedserie*			
Constant Con			
Control dependency algo			

799 views

🖆 9 🐠 1 🏕 SHARE ≕+ •••





Public Feedback: Model Visualization is a key feature of TensorFlow

Public Feedback: Model Visualization is a key feature of TensorFlow

"One of the main lacking areas of almost all open source Machine Learning packages, was the ability to visualize model and follow the computation pipeline" - Quora

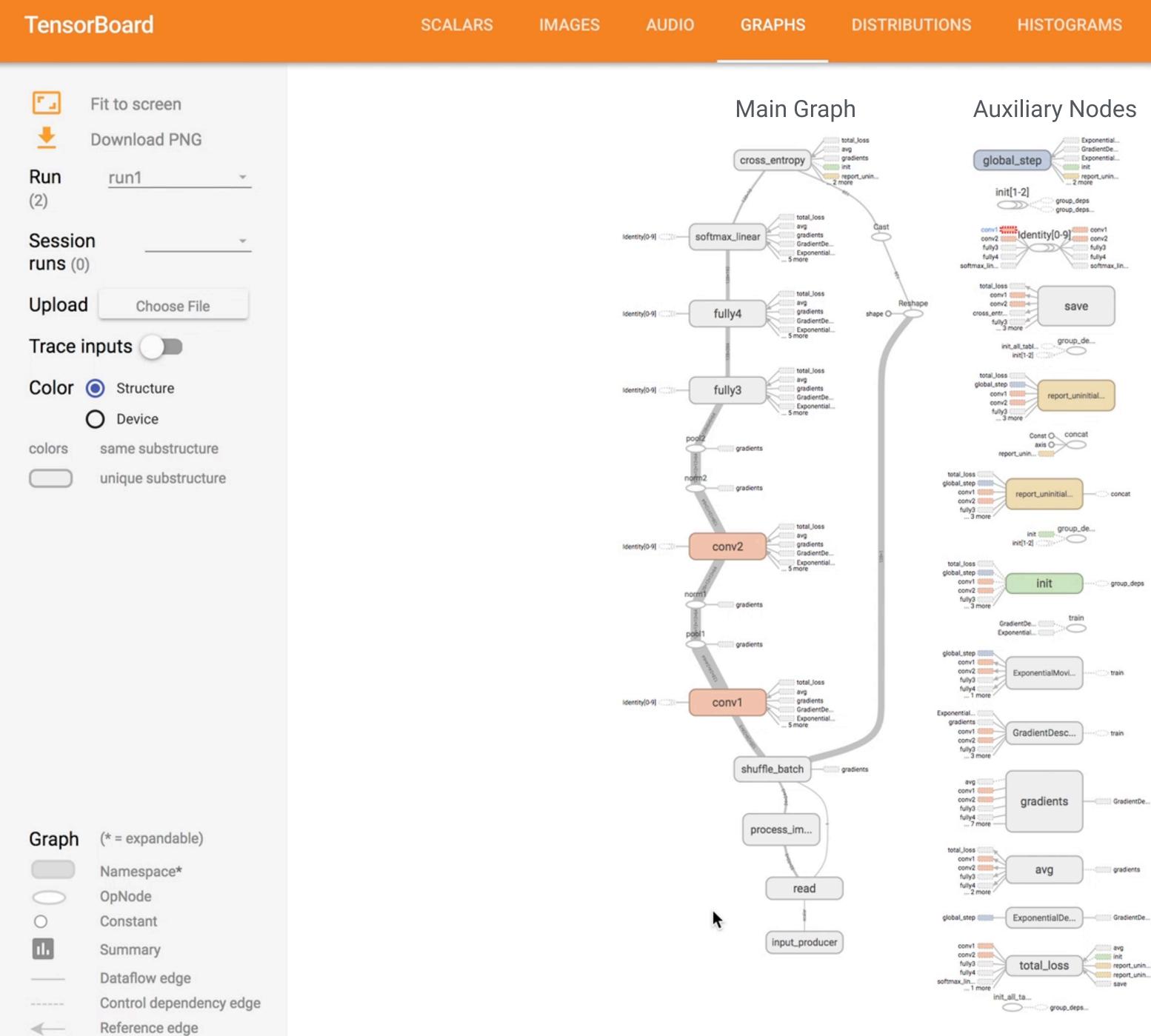
Public Feedback: Model Visualization is a key feature of TensorFlow

"One of the main lacking areas of almost all open source Machine Learning packages, was the ability to visualize model and follow the computation pipeline" - Quora

"We believe visualization is really fundamental to the creative process and our ability to develop better models. So, visualization tools like TensorBoard are a great step in the right direction." – Indico



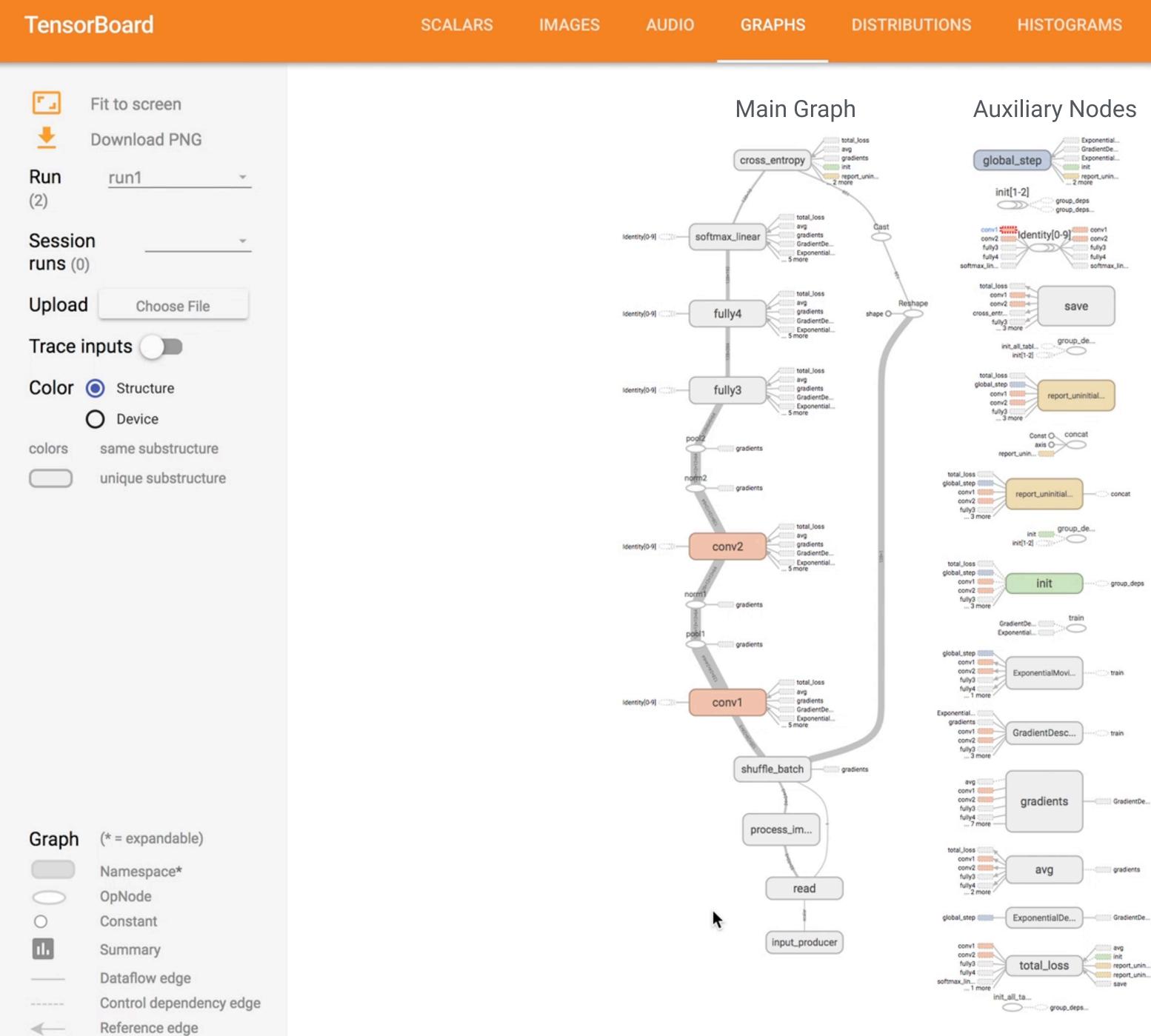






-

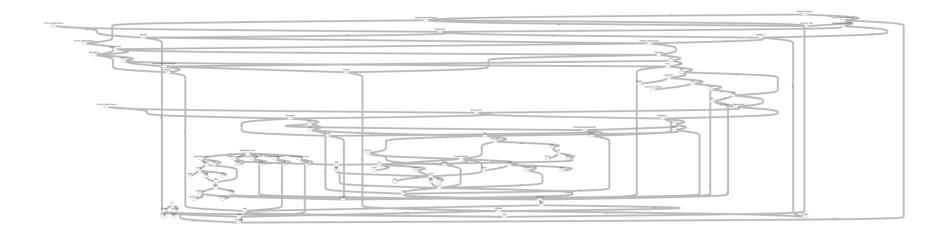
?



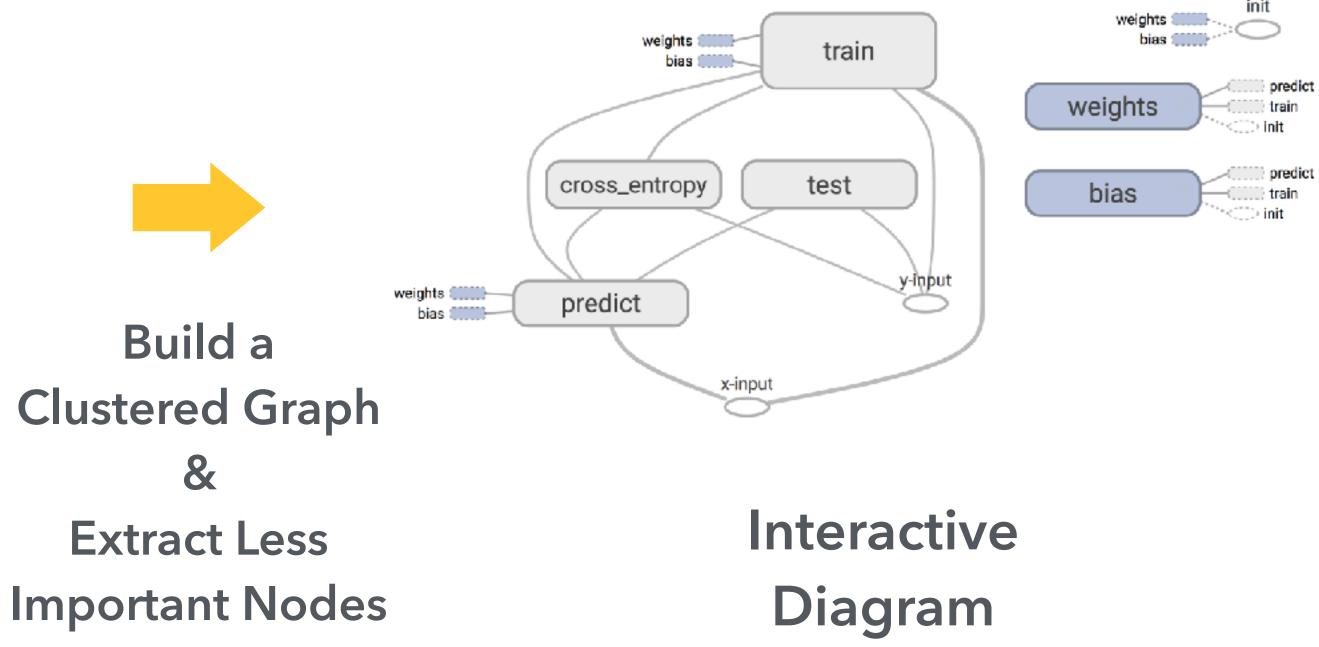


-

?

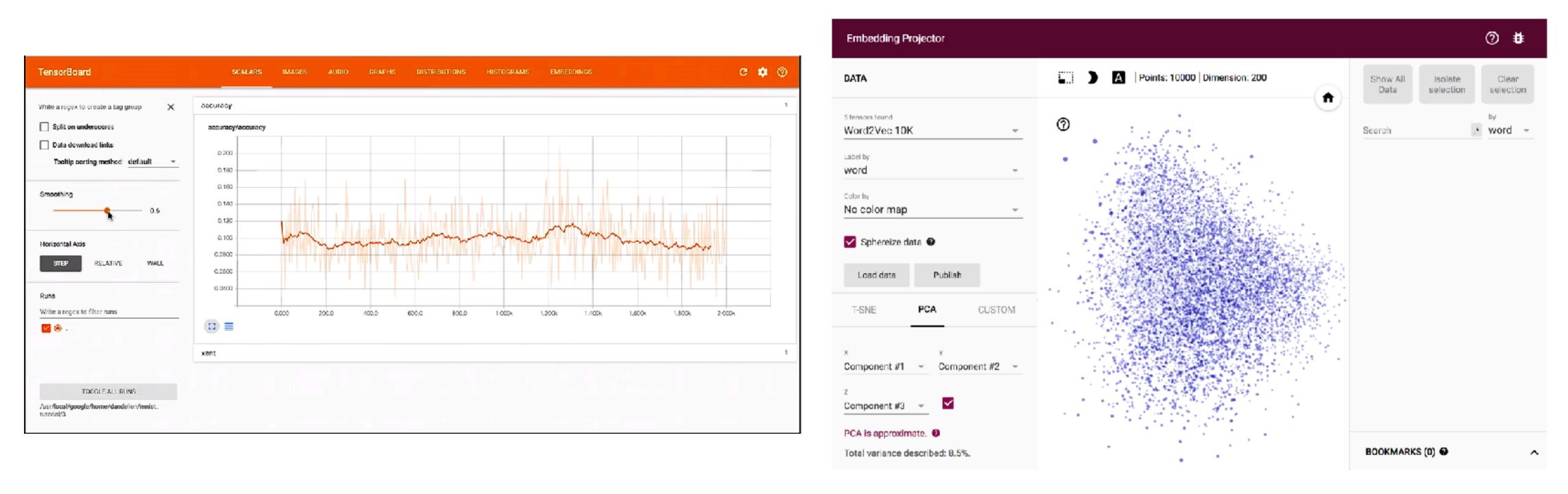


Low-level **Dataflow Graph**





Visualization can play many important roles for machine learning



This repository Search	Pu	ll requests Issues	Marketplace	Explore				\$ -
tensorflow / tensorboard	I			💿 Watch 🗸	54	🛨 Star	460	Ϋ́Fo
<> Code () Issues (133)	13 Pull requests	Insights -						
ensorFlow's Visualization Too	olkit							
1,653 commits	រ៉ូវ 12 branches	\heartsuit 5 releas	ses	22 89 cont	ributors		ৰাই Apa	ache-2
Branch: master - New pull reque	st		Create	new file Up	load files	Find file	Clor	ne or de
piperchester committed with cl	h ihuahua Fix code forma	tting example in READ	ME (#602) ···			Latest con	nmit a058	Bcb5 8
tensorboard	Remove the hatch	hed background from	n functions (#60	01)				8 h
third_party	Fix broken Polym	er URL (#594)						2
.gitignore	Ignore pyc and eg	gg files (#105)						3 mo
J.pylintrc	Add progress bar	to loader (#200)						3 mo
.travis.yml	Make TensorBoar	d work with latest Ba	azel (#590)					2
ADDING_A_PLUGIN.md	Link to docs withi	in tensorboard-plugi	n-example (#41	2)				a m
	Add AUTHORS, L	ICENSE						4 mo
	Make CONTRIBU	TING.md consistent	with TensorFlow	v (#580)				3
DEVELOPMENT.md	Note how we dev	elop on TF nightly (#	486)					25
								1
	Add AUTHORS, L	ICENSE						4 mo
E LICENSE		ICENSE ng example in READN	ME (#602)					4 mo 8 h

https://github.com/tensorflow/tensorboard



e-2.0

download -

8 hours ago

3 hours ago

2 days ago

months ago

nonths ago

2 days ago

month ago

months ago

3 days ago

25 days ago

months ago

8 hours ago

2 days ago

This repository Search	Pu	ll requests Issues	Marketplace	Explore				\$ -
tensorflow / tensorboard	I			💿 Watch 🗸	54	🛨 Star	460	Ϋ́Fo
<> Code () Issues (133)	13 Pull requests	Insights -						
ensorFlow's Visualization Too	olkit							
1,653 commits	រ៉ូវ 12 branches	\heartsuit 5 releas	ses	22 89 cont	ributors		ৰাই Apa	ache-2
Branch: master - New pull reque	st		Create	new file Up	load files	Find file	Clor	ne or de
piperchester committed with cl	h ihuahua Fix code forma	tting example in READ	ME (#602) ···			Latest con	nmit a058	Bcb5 8
tensorboard	Remove the hatch	hed background from	n functions (#60	01)				8 h
third_party	Fix broken Polym	er URL (#594)						2
.gitignore	Ignore pyc and eg	gg files (#105)						3 mo
J.pylintrc	Add progress bar	to loader (#200)						3 mo
.travis.yml	Make TensorBoar	d work with latest Ba	azel (#590)					2
ADDING_A_PLUGIN.md	Link to docs withi	in tensorboard-plugi	n-example (#41	2)				a m
	Add AUTHORS, L	ICENSE						4 mo
	Make CONTRIBU	TING.md consistent	with TensorFlow	v (#580)				3
DEVELOPMENT.md	Note how we dev	elop on TF nightly (#	486)					25
								1
	Add AUTHORS, L	ICENSE						4 mo
E LICENSE		ICENSE ng example in READN	ME (#602)					4 mo 8 h

https://github.com/tensorflow/tensorboard



e-2.0

download -

8 hours ago 2 days ago 2 days ago nonths ago 2 days ago 2 days ago 2 days ago months ago 3 days ago 5 days ago 5 days ago

2 days ago

Build your own Machine Learning Visualizations with the new TensorBoard API

Monday, September 11, 2017

Posted by Chi Zeng and Justine Tunney, Software Engineers, Google Brain Team

When we open-sourced TensorFlow in 2015, it included TensorBoard, a suite of visualizations for inspecting and understanding your TensorFlow models and runs. Tensorboard included a small, predetermined set of visualizations that are generic and applicable to nearly all deep learning applications such as observing how loss changes over time or exploring clusters in highdimensional spaces. However, in the absence of reusable APIs, adding new visualizations to TensorBoard was prohibitively difficult for anyone outside of the TensorFlow team, leaving out a long tail of potentially creative, beautiful and useful visualizations that could be built by the research community.

To allow the creation of new and useful visualizations, we announce the release of a consistent set of APIs that allows developers to add custom visualization plugins to TensorBoard. We hope that developers use this API to extend TensorBoard and ensure that it covers a wider variety of use cases.

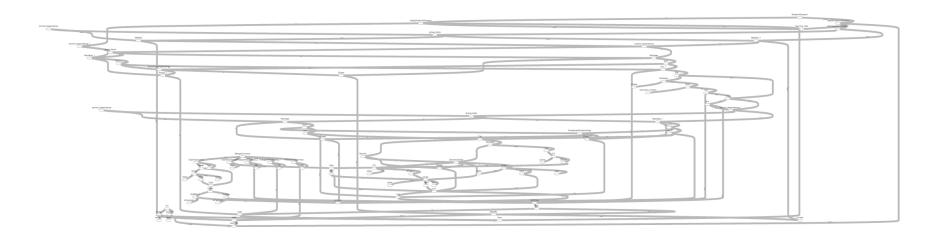
We have updated the existing dashboards (tabs) in TensorBoard to use the new API, so they serve as examples for plugin creators. For the current listing of plugins included within TensorBoard, you can explore the tensorboard/plugins directory on GitHub. For instance, observe the new plugin that generates precision-recall curves:





Visualizing Dataflow Graphs of **Deep Learning Models in TensorFlow**

Kanit "Ham" Wongsuphasawat, Daniel Smilkov, James Wexler, Jimbo Wilson, Dandelion Mané, Doug Fritz, Dilip Krishnan, Fernanda B. Viégas, Martin Wattenberg



Low-level **Dataflow Graph**

